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A Note on Integer Linear Programming and Incidence Treedepth

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Recently a strong connection has been shown between tractability of integer programming (IP) with bounded coefficients on the one side and the structure of the primal or dual Gaifmann graph on the other side. To that end, integer linear programming is fixed-parameter tractable with respect to the primal (or dual) treedepth of its constraint matrix and the largest coefficient (in absolute value). Here, primal and dual treedepth refer to the treedepth of the primal and dual Gaifman graph (of the constraint matrix). Motivated by this, Koutecký, Levin, and Onn [ICALP 2018] asked whether it is possible to extend these result to a more broader class of integer linear programs. More formally, is integer linear programming fixed-parameter tractable with respect to the incidence treedepth of its constraint matrix and the largest coefficient (in absolute value)?

We answer this question in negative. We prove that deciding the feasibility of a system in the standard form, $A\mathbf{x} = \mathbf{b}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}$, is NP-hard even when the absolute value of any coefficient in A is 1 and the incidence treedepth of A is 5. Consequently, it is not possible to decide feasibility in polynomial time even if both the assumed parameters are constant, unless $P = NP$.

Appendix

1 Introduction

In this paper we consider the decision version of Integer Linear Program (ILP) in *standard form*. Here, given a matrix $A \in \mathbb{Z}^{m \times n}$ with m rows (constraints) and n columns and vectors $\mathbf{b} \in \mathbb{Z}^m$ and $\mathbf{l}, \mathbf{u} \in \mathbb{Z}^n$ the task is to decide whether the set

$$\{\mathbf{x} \in \mathbb{Z}^n \mid A\mathbf{x} = \mathbf{b}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}\} \quad (\text{SSol})$$

is non-empty. We are going to study structural properties of the incidence graph of the matrix A . An integer program (IP) is a *standard IP* (SIP) if its set of solutions is

described by (??), that is, if it is of the form

$$\min \{f(\mathbf{x}) \mid A\mathbf{x} = \mathbf{b}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}, \mathbf{x} \in \mathbb{Z}^n\}, \quad (\text{SIP})$$

where $f: \mathbb{N}^n \rightarrow \mathbb{N}$ is the *objective function*; in case f is a linear function the above SIP is said to be a linear SIP. Before we go into more details we first review some recent development concerning algorithms for solving (linear) SIPs in variable dimension with the matrix A admitting a certain decomposition.

Let E be a 2×2 block matrix, that is, $E = \begin{pmatrix} A_1 & A_2 \\ A_3 & A_4 \end{pmatrix}$, where A_1, \dots, A_4 are integral matrices. We define an *n -fold 4-block product of E* for a positive integer n as the following block matrix

$$E^{(n)} = \begin{pmatrix} A_1 & A_2 & A_2 & \cdots & A_2 \\ A_3 & A_4 & 0 & \cdots & 0 \\ A_3 & 0 & A_4 & \cdots & 0 \\ \vdots & & & \ddots & \\ A_3 & 0 & 0 & \cdots & A_4 \end{pmatrix},$$

where 0 is a matrix containing only zeros (of appropriate size). One can ask whether replacing A in the definition of the set of feasible solutions (??) can give us an algorithmic advantage leading to an efficient algorithm for solving such SIPs. We call such an SIP an *n -fold 4-block IP*. We derive two special cases of the n -fold 4-block IP with respect to special cases for the matrix E (see monographs [DeLoeraHK13; Onn10] for more information). If both A_1 and A_3 are void (not present at all), then the result of replacing A with $E^{(n)}$ in (??) yields the *n -fold IP*. Similarly, if A_1 and A_2 are void, we obtain the *2-stage stochastic IP*.

The first, up to our knowledge, pioneering algorithmic work on n -fold 4-block IPs is due to Hemmecke et al. [HemmeckeKW10]. They gave an algorithm that given n , the 2×2 block matrix E , and vectors $\mathbf{w}, \mathbf{b}, \mathbf{l}, \mathbf{u}$ finds an integral vector \mathbf{x} with $E^{(n)}\mathbf{x} = \mathbf{b}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}$ minimizing $\mathbf{w}\mathbf{x}$. The algorithm of Hemmecke et al. [HemmeckeKW10] runs in time $n^{g(r,s,\|E\|_\infty)}L$, where r is the number of rows of E , s is the number of columns of E , L is the size of the input, and $g: \mathbb{N} \rightarrow \mathbb{N}$ is a computable function. Thus, from the parameterized complexity viewpoint this is an XP algorithm for parameters $r, s, \|E\|_\infty$. This algorithm has been recently improved by Chen et al. [ChenXS18] who give better bounds on the function g ; it is worth noting that Chen et al. [ChenXS18] study also the special case where A_1 is a zero matrix and even in that case present an XP algorithm. Since the work of Hemmecke et al. [HemmeckeKW10] the question of whether it is possible to improve the algorithm to run in time $g'(r, s, \|E\|_\infty) \cdot n^{O(1)}L$ or not has become a major open question in the area of mathematical programming.

Of course, the complexity of the two aforementioned special cases of n -fold 4-block IP are extensively studied as well. The first FPT algorithm¹ for the n -fold IPs (for parameters $r, s, \|E\|_\infty$) is due to Hemmecke et al. [HemmeckeOR13]. Their algorithm has been subsequently improved [KouteckyLO18; EisenbrandHK18]. Altmanová et al. [AltmanovaKK18] implemented the algorithm of Hemmecke et al. [HemmeckeOR13]

¹That is an algorithm running in time $f(r, s, \|E\|_\infty) \cdot n^{O(1)}L$.

and improved the polynomial factor (achieving the same running time as Eisenbrand et al. [EisenbrandHK18]) the above algorithms (from cubic dependence to $n^2 \log n$). The best running time of an algorithm solving n -fold IP is due to Jansen et al. [JansenLR18] and runs in nearly linear time in terms of n .

Last but not least, there is an FPT algorithm for solving the 2-stage stochastic IP due to Hemmecke and Schultz [HemmeckeS03]. This algorithm is, however, based on a well quasi ordering argument yielding a bound on the size of the Graver basis for these IPs. Very recently Klein [Klein19] presented a constructive approach using Steinitz lemma and give the first explicit (and seemingly optimal) bound on the size of the Graver basis for 2-stage (and multistage) IPs. It is worth noting that possible applications of 2-stage stochastic IP are much less understood than those of its counterpart n -fold IP.

In the past few years, algorithmic research in this area has been mainly application-driven. Substantial effort has been taken in order to find the right formalism that is easier to understand and yields algorithms having the best possible ratio between their generality and the achieved running time. It turned out that the right formalism is connected with variants of the Gaifman graph (see e.g. [Dechter06]) of the matrix A (for the definitions see the Preliminaries section).

Our Contribution. In this paper we focus on the incidence (Gaifman) graph. We investigate the (negative) effect of the treedepth of the incidence Gaifman graph on tractability of ILP feasibility.

Theorem 1. *Given a matrix $A \in \{-1, 0, 1\}^{m \times n}$ and vectors $\mathbf{l}, \mathbf{u} \in \mathbb{Z}_{\infty}^n$. Deciding whether the set defined by (??) is non-empty is NP-hard even if $\mathbf{b} = \mathbf{0}$ and $\text{td}_I(A) \leq 5$.*

Preliminaries

For integers $m < n$ by $[m : n]$ we denote the set $\{m, m + 1, \dots, n\}$ and $[n]$ is a shorthand for $[1 : n]$. We use bold face letters for vectors and normal font when referring to their components, that is, \mathbf{x} is a vector and x_3 is its third component. For vectors of vectors we first use superscripts to access the “inner vectors”, that is, $\mathbf{x} = (\mathbf{x}^1, \dots, \mathbf{x}^n)$ is a vector of vectors and \mathbf{x}^3 is the third vector in this collection.

From Matrices to Graphs. Let A be an $m \times n$ integer matrix. The *incidence Gaifman graph* of A is the bipartite graph $G_I = (R \cup C, E)$, where $R = \{r_1, \dots, r_m\}$ contains one vertex for each row of A and $C = \{c_1, \dots, c_n\}$ contains one vertex for each column of A . There is an edge $\{r, c\}$ between the vertex $r \in R$ and $c \in C$ if $A(r, c) \neq 0$, that is, if row r contains a nonzero coefficient in column c . The *primal Gaifman graph* of A is the graph $G_P = (C, E)$, where C is the set of columns of A and $\{c, c'\} \in E$ whenever there exists a row of A with a nonzero coefficient in both columns c and c' . The *dual Gaifman graph* of A is the graph $G_D = (R, E)$, where R is the set of rows of A and $\{r, r'\} \in E$ whenever there exists a column of A with a nonzero coefficient in both rows r and r' .

Treedepth. Undoubtedly, the most celebrated structural parameter for graphs is treewidth, however, in the case of ILPs bounding treewidth of any of the graphs defined above does not lead to tractability (even if the largest coefficient in A is bounded as well see e.g. [KouteckyLO18]). Treedepth is a structural parameter well suited for the theory of so called sparse graph classes [NesetrilOdM:Sprarsity]². Let $G = (V, E)$ be a graph. The treedepth of G , denoted $\text{td}(G)$, is defined by the following formula:

$$\text{td}(G) = \begin{cases} 1 & \text{if } |V(G)| = 1, \\ 1 + \min_{v \in V(G)} \text{td}(G - v) & \text{if } G \text{ is connected with } |V(G)| > 1, \\ \max_{i \in [k]} \text{td}(G_i) & \text{if } G_1, \dots, G_k \text{ are connected components of } G. \end{cases}$$

An example was asked by Rew #2; it might be helpful to readers from the IP-theory community. Do we have any good example?

Let A be an $m \times n$ integer matrix. The *incidence treedepth* of A , denoted $\text{td}_I(A)$, is the treedepth of its incidence Gaifman graph G_I . The *dual treedepth* of A , denoted $\text{td}_D(A)$, is the treedepth of its dual Gaifman graph G_D . The *primal treedepth* is defined similarly.

The following two well-known theorems will be used in the proof of Theorem ??.

Theorem 2 (Chinese Remainder Theorem). *Let p_1, \dots, p_n be pairwise co-prime integers greater than 1 and let a_1, \dots, a_n be integers such that for all $i \in [n]$ it holds $0 \leq a_i < p_i$. Then there exists exactly one integer x such that*

1. $0 \leq x < \prod_{i=1}^n p_i$ and
2. $\forall i \in [n] : x \equiv a_i \pmod{p_i}$.

Theorem 3 (Prime Number Theorem). *Let $\pi(n)$ denote the number of primes in $[n]$, then $\pi(n) \sim \frac{n}{\log(n)}$.*

It is worth pointing out that, given a positive integer n encoded in unary, it is possible to the n -th prime in polynomial time.

2 Proof of Theorem ??

Before we proceed to the proof of Theorem ?? we include a brief sketch of its idea. To prove NP-hardness, we will give a polynomial time reduction from 3-SAT which is well known to be NP-complete [GareyJ79]. The proof is inspired by the NP-hardness proof for ILPs given by a set of inequalities, where the primal graph is a star, of Eiben et. al [EibenGKO18].

²See also lecture notes available at <https://www.mimuw.edu.pl/~mp248287/sparsity/>

Proof Idea. Let φ be a 3-CNF formula. We encode an assignment into a variable y . With every variable v_i of the formula φ we associate a prime number p_i and let $y \bmod p_i$ be the boolean value of the variable v_i . Furthermore, if for a clause $C \in \varphi$ by $\|C\|$ we denote the product of all of the primes associated with the variables occurring in C . Then, by Chinese Remainder Theorem, there is a single value in $[\|C\|]$, associated with the assignment that falsifies C , which we have to forbid for $y \bmod \|C\|$ (we use the box constraints, i.e., the vectors \mathbf{l}, \mathbf{u} , for an auxiliary variable taking the value $y \bmod \|C\|$ to achieve this). For example let $\varphi = (v_1 \vee \neg v_2 \vee v_3)$ and let the primes associated with the three variables be 2, 3, and 5, respectively. Then we have $\|(v_1 \vee \neg v_2 \vee v_3)\| = 30$ and, since $v_1 = v_3 = \mathbf{false}$ and $v_2 = \mathbf{true}$ is the only assignment falsifying this clause, we have that 21 is the forbidden value for $y \bmod 30$. It is worth pointing out that for a clause C there is exactly one number in $\|C\|$ yielding the falsifying assignment to the three variables in C . Finally, the (??) constructed from φ is feasible if and only if there is a satisfying assignment for φ .

Proof of Theorem ??. Let φ be a 3-CNF formula with n' variables $v_1, \dots, v_{n'}$ and m' clauses $C_1, \dots, C_{m'}$ (an instance of 3-SAT). Note that we can assume that none of the clauses in φ contains a variable along with its negation. We will define an SIP, that is, vectors $\mathbf{b}, \mathbf{l}, \mathbf{u}$, and a matrix A with $\mathcal{O}((n' + m')^5)$ rows and columns, whose solution set is non-empty if and only if a satisfying assignment exists for φ . Furthermore, we present a decomposition of the incidence graph of the constructed SIP proving that its treedepth is at most 5. We naturally split the vector \mathbf{x} of the SIP into subvectors associated with the sought satisfying assignment, variables, and clauses of φ , that is, we have $\mathbf{x} = (y, \mathbf{x}^1, \dots, \mathbf{x}^{n'}, \mathbf{z}^1, \dots, \mathbf{z}^{m'})$. Throughout the proof p_i denotes the i -th prime number.

Variable Gadget. We associate the $\mathbf{x}^i = (x_0^i, \dots, x_{p_i}^i)$ part of \mathbf{x} with the variable v_i and bind the assignment of v_i to y . We add the following constraints

$$x_1^i = x_\ell^i \quad \forall \ell \in [2 : p_i] \quad (1)$$

$$x_0^i = y + \sum_{\ell=1}^{p_i} x_\ell^i \quad (2)$$

and box constraints

$$-\infty \leq x_\ell^i \leq \infty \quad \forall \ell \in [p_i] \quad (3)$$

$$0 \leq x_0^i \leq 1 \quad (4)$$

to the SIP constructed so far.

Claim 1. *An integral vector \mathbf{x} satisfies (??) and (??) if and only if $x_0^i \equiv y \bmod p_i$.*

Proof. By (??) we know $x_1^i = \dots = x_{p_i}^i$ and thus by substitution we get the following equivalent form of (??)

$$x_0^i = y + p_i \cdot x_1^i. \quad (5)$$

But this form is equivalent to $x_0^i \equiv y \pmod{p_i}$ for an integral vector \mathbf{x} . \lrcorner

Note that by (the proof of) the above claim the conditions (??) and (??) essentially replace the large coefficient (p_i) used in the condition (??). This is an efficient trade-off between large coefficients and incidence treedepth which we are going to exploit once more when designing the clause gadget.

By the above claim we get an immediate correspondence between y and truth assignments for $v_1, \dots, v_{n'}$. For an integer w and a variable v_i we define the following mapping

$$(w, v_i) = \begin{cases} \mathbf{true} & \text{if } w \equiv 1 \pmod{p_i} \\ \mathbf{false} & \text{if } w \equiv 0 \pmod{p_i} \\ \text{undefined} & \text{otherwise.} \end{cases}$$

Notice that (??) implies that the the mapping $(y, v_i) \in \{\mathbf{true}, \mathbf{false}\}$ for $i \in [n']$. We straightforwardly extend the mapping (\cdot, \cdot) for tuples of variables as follows. For a tuple \mathbf{a} of length ℓ , the value of (w, \mathbf{a}) is $((w, a_1), \dots, (w, a_\ell))$ and we say that (w, \mathbf{a}) is defined if all of its components are defined.

Clause Gadget. Let C_j be a clause with variables v_e, v_f, v_g . We define $\|C_j\|$ as the product of the primes associated with the variables occurring in C_j , that is, $\|C_j\| = p_e \cdot p_f \cdot p_g$. We associate the $\mathbf{z}^j = (z_0^j, \dots, z_{\|C_j\|}^j)$ part of \mathbf{x} with the clause C_j . Let d_j be the unique integer in $[\|C_j\|]$ for which $(d_j, (v_e, v_f, v_g))$ is defined and gives the falsifying assignment for C_j . The existence and uniqueness of d_j follows directly from the Chinese Remainder Theorem. We add the following constraints

$$z_1^j = z_\ell^j \quad \forall \ell \in [2 : \|C_j\|] \quad (6)$$

$$z_0^j = y + \sum_{1 \leq \ell \leq \|C_j\|} z_\ell^j \quad (7)$$

and box constraints

$$-\infty \leq z_\ell^j \leq \infty \quad \forall \ell \in [\|C_j\|] \quad (8)$$

$$d_j + 1 \leq z_0^j \leq \|C_j\| + d_j - 1 \quad (9)$$

to the SIP constructed so far.

Claim 2. *Let C_j be a clause in φ with variables v_e, v_f, v_g . An integral vector \mathbf{x} satisfies (??), (??), (??), (??), (??), and (??) if and only if $(y, (v_e, v_f, v_g))$ is defined and satisfies C_j .*

Proof. We have already seen that the first claim together with (??) is equivalent to $(y, (v_e, v_f, v_g))$ is defined. Now, similarly to the proof of the first claim, (??) and (??) together are equivalent to $z_0^j \equiv y \pmod{\|C_j\|}$. Finally, by (??) we obtain that $z_0^j \neq d_j$ which holds if and only if $(y, (v_e, v_f, v_g))$ satisfies C_j . \lrcorner

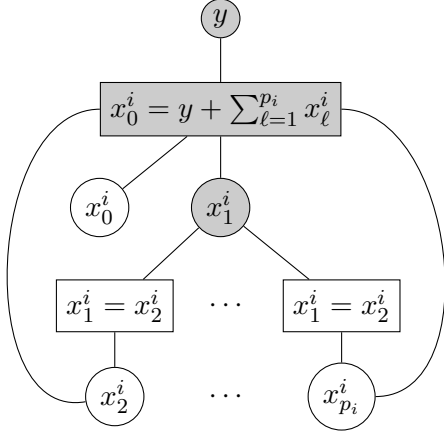


Figure 1: The variable gadget for u_i of 3-SAT instance together with the global variable y . Variables (of the IP) are in circular nodes while equations are in rectangular ones. The nodes deleted in the proof of the third claim in the proof of Theorem ?? have light gray background.

Let $A\mathbf{x} = \mathbf{0}$ be the SIP with constraints (??), (??), (??), and (??) and box constraints $\mathbf{l} \leq \mathbf{x} \leq \mathbf{u}$ given by (??), (??), (??), (??), and $-\infty \leq y \leq \infty$. By the two claims above we have that $A\mathbf{x} = \mathbf{0}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}$ is feasible if and only if $(y, (v_1, \dots, v_{n'}))$ is defined and satisfies every clause in φ . This finishes the reduction.

In order to finish the proof we have to bound the number of variables and constraints in the presented SIP and to bound the incidence treedepth of A . It follows from the Prime Number Theorem that $p_i = \mathcal{O}(i \log(i))$. Hence, the number of rows and columns of A is at most $(n' + m')p_{n'}^3 = \mathcal{O}((n' + m')^5)$.

Claim 3. *It holds that $\text{td}_I(A) \leq 5$.*

Proof. Let G be the incidence graph of the matrix A . It is easy to verify that y is a cut-vertex in G . Observe that each component of $G - y$ is now either a variable gadget for v_i with $i \in [n']$ (we call such a component a *variable component*) or a clause gadget for C_j with $j \in [m']$ (we call such a component a *clause component*). Let G_v^i be the variable component (of $G - y$) containing variables \mathbf{x}^i and G_c^j be the clause component containing variables \mathbf{z}^j . Let $t_v = \max_{\ell \in [n']} \text{td}(G_v^\ell)$ and $t_c = \max_{\ell \in [m']} \text{td}(G_c^\ell)$. It follows that $\text{td}(G) \leq 1 + \max(t_v, t_c)$.

Refer to Figure ??. Observe that if we delete the variable x_1^i together with the constraint (??) from G_v^i , then each component in the resulting graph contains at most two vertices. Each of these components contains either

- a variable x_ℓ^i and an appropriate constraint (??) (the one containing x_ℓ^i and x_0^i) for some $\ell \in [2 : p_i]$ or
- the variable x_0^i .

Since treedepth of an edge is 2 and treedepth of the one vertex graph is 1, we have that $t_v \leq 4$.

The bound on t_c follows the same lines as for t_v , since indeed the two gadgets have the same structure. Now, after deleting z_1^j and (??) in G_c^j we arrive to a graph with treedepth of all of its components again bounded by two (in fact, none of its components contain more than two vertices). Thus, $t_v \leq 4$ and the claim follows. \lrcorner

The theorem follows by combining the three above claims. \square

3 Incidence Treedepth of Restricted ILPs

It is worth noting that the proof of Theorem ?? crucially relies on having variables as well as constraints which have high degree in the incidence graph. Thus, it is natural to ask whether this is necessary or, equivalently, whether bounding the degree of variables, constraints, or both leads to tractability. It is well known that if a graph G has bounded degree and treedepth, then it is of bounded size, since indeed the underlying decomposition tree has bounded height and degree and thus bounded number of vertices. Let (??) with n variables be given. Let $c(A)$ denote the maximum arity of a constraint in its constraint matrix A and let $v(A)$ denote the maximum occurrence of a variable in constraints of A . In other words, $c(A)$ denotes the maximum number of nonzeros in a row of A and $v(A)$ denotes the maximum number of nonzeros in a column of A . Now, we get that ILP can be solved in time $f(c(A), v(A), \text{td}_I(A))L^{O(1)}$, where f is some computable function and L is the length of the encoding of the given ILP thanks to Lenstra's algorithm [Lenstra83].

The above observation can in fact be strengthened—namely, if the arity of all the constraints or the number of occurrences of all the variables in the given SIP is bounded, then we obtain a bound on either primal or dual treedepth. This is formalized by the following technical lemma.

Lemma 4. *For every (??) we have $\text{td}_P(A) \leq c \cdot \text{td}_I(A)$ and $\text{td}_D(A) \leq v \cdot \text{td}_I(A)$.*

The proof idea is to exploit certain properties of the decomposition of the incidence graph of A and turn it into a decomposition of the primal or the dual graph of A . To do so we replace the nodes corresponding to variables (or constraints, respectively) of it with a path on v (c) nodes. Finally, it is shown that one can find a suitable subtree of the resulting tree yielding a valid decomposition for the dual or primal graph of A . In order to do so we have to move some vertices closer to the root of the new decomposition; roughly speaking we move a variable to some node of the path we have used to replace the constraint that contains the variable and is the closest such to the root (the construction for constraints is similar).

Proof of Lemma ??. We show how to transform a decomposition of the incidence graph of A into a decomposition of the dual graph of A showing the claimed bound. We note that the proof for the primal graph is analogous and follows from a symmetric argument (take a transpose of A and apply the same procedure).

Let G_I be the incidence graph of A and let G_D be the dual graph of A . Let T be a rooted tree with depth d and let $\iota: V(G_I) \rightarrow V(T)$ be the mapping that together with T shows $\text{td}_I(A) \leq d$. We show that there exists a rooted tree \hat{T} of depth at most $d \cdot v$ and a mapping $\delta: V(G_D) \rightarrow V(\hat{T})$ such that there is a root to leaf path containing vertices $\delta(v), \delta(u)$ for each edge $uv \in E(G_D)$. In order to improve readability we refer to vertices of the trees (e.g. T) as to *nodes*.

We split $V(G_I)$ into two sets $V_V \dot{\cup} V_C$ in such a way that all vertices corresponding to a variable of the given SIP are in the set V_V (and for constraints we have V_C). Note

that $V(G_D) = V_C$. We begin by transforming T into a new tree T' in which we replace each node v for which the vertex $\iota^{-1}(v) \in V_V$ by a path with ν nodes; let V_v denote the set of vertices of this path. Now, we define a mapping $\gamma: V_C \rightarrow V_V$ from constraints to variables by $\gamma(c) = v$ if v is closest to the root of T with $cv \in E(G_I)$ (for $c \in V_C$). Now we are ready to define an injective mapping $\delta': V_C \rightarrow V(\hat{T})$. For a vertex $c \in V_C$ we define $\delta'(c)$ to be any so far unused vertex in $V_{\gamma(c)}$. This is always possible, since we have $|N_{G_I}(c)| \leq_C = |V_{\gamma(c)}|$. Now, we contract all edges of T' for which δ' maps no vertex of V_C to at least one of its endpoints; let \hat{T} be the tree obtained after all of the contractions. Let $\delta: V_C \rightarrow V(\hat{T})$ denote the restriction of the mapping δ' (in fact, this is the same mapping, since we have preserved all nodes to which δ' maps some vertex of V_C). We claim that \hat{T} and δ form a valid treedepth decomposition for G_D . In order to see this note that $cc' \in E(G_D)$ if and only if there exists a variable that occurs in both c and c' . Let v be any such variable. Now, by the definition of δ both $\delta(c), \delta(c')$ must in T' be placed on the path between root and the bottom node of the path replacing v in T . Consequently, $\delta(c)$ and $\delta(c')$ are in parent-ancestor relation defined by \hat{T} . Observe that, since there are at most $\text{td}_I(A)$ nodes on each root to leaf path in T , there are at most $\nu \cdot \text{td}_I(A)$ nodes on each root to leaf path in T' as well as in \hat{T} . \square

It follows that if we bound either $\nu(A)$ or $C(A)$, that is, formally set $= \min\{\nu(A), C(A)\}$, then the linear IP with such a solution set is solvable in time $f(\cdot, \|A\|_\infty) \cdot n^{O(1)} \cdot L$ thanks to results of Koutecký et al. [KouteckýLO18]. Consequently, the use of high-degree constraints and variables in the proof of Theorem ?? is unavoidable.

4 Conclusions

We have shown that, unlike the primal and the dual treedepth, the incidence treedepth of a constraint matrix of (??) does not (together with the largest coefficient) provide a way to tractability. This shows our current understanding of the structure of the incidence Gaifman graph is not sufficient. Thus, the effect on tractability of some other “classical” graph parameters shall be investigated. For example we have some preliminary evidences that

- the vertex cover number of the incidence Gaifman graph together with the largest coefficient yields a tractable case and
- the graph in our reduction (Theorem ??) may admit a treecut decomposition of constant width.

We are going to investigate the two above claims in detail in the full version of this paper. Last but not least, all of the above suggest some open questions. Namely, whether ILP parameterized by the largest coefficient and treewidth and the maximum degree of the incidence Gaifman graph is in FPT or not. Furthermore, one may also ask about parameterization by the largest coefficient and the feedback vertex number of the incidence Gaifman graph.