

A Polynomial Degree Bound on Equations for Non-Rigid Matrices and Small Linear Circuits

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Abstract

We show that there is an equation of degree at most $\text{poly}(n)$ for the (Zariski closure of the) set of the non-rigid matrices: that is, we show that for every large enough field \mathbb{F} , there is a non-zero n^2 -variate polynomial $P \in \mathbb{F}[x_{1,1}, \dots, x_{n,n}]$ of degree at most $\text{poly}(n)$ such that every matrix M which can be written as a sum of a matrix of rank at most $n/100$ and a matrix of sparsity at most $n^2/100$ satisfies $P(M) = 0$. This confirms a conjecture of Gesmundo, Hauenstein, Ikenmeyer and Landsberg [8] and improves the best upper bound known for this problem down from $\exp(n^2)$ [11, 8] to $\text{poly}(n)$.

We also show a similar polynomial degree bound for the (Zariski closure of the) set of all matrices M such that the linear transformation represented by M can be computed by an algebraic circuit with at most $n^2/200$ edges (without any restriction on the depth). As far as we are aware, no such bound was known prior to this work when the depth of the circuits is unbounded.

Our methods are elementary and short and rely on a polynomial map of Shpilka and Volkovich [17] to construct low degree “universal” maps for non-rigid matrices and small linear circuits. Combining this construction with a simple dimension counting argument to show that any such polynomial map has a low degree annihilating polynomial completes the proof.

As a corollary, we show that any derandomization of the polynomial identity testing problem will imply new circuit lower bounds. A similar (but incomparable) theorem was proved by Kabanets and Impagliazzo [10].

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1 Introduction

1.1 Equations for varieties in algebraic complexity theory

Let $V \subseteq \mathbb{F}^n$ be a (not necessarily irreducible) affine variety and let $\mathbf{I}(V)$ denote its ideal.² A non-zero polynomial $P \in \mathbf{I}(V)$ is called an *equation* for V . An equation for V may serve as a “proof” that a point $\mathbf{x} \in \mathbb{F}^n$ is *not* in V , by showing that $P(\mathbf{x}) \neq 0$.

A fundamental observation of the Geometric Complexity Theory program is that many important circuit lower bounds problems in algebraic complexity theory fit naturally into the setting of showing that a point \mathbf{x} lies outside a variety V [15, 5]. In this formulation, one considers V to be the closure of a class of polynomials of low complexity, and \mathbf{x} is the coefficient vector of the candidate hard polynomial.

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² For completeness, we provide the formal (standard) definitions for these notions in Subsection 1.4.



Let $\Delta(V) := \min_{0 \neq P \in \mathcal{I}(V)} \{\deg(P)\}$. The quantity $\Delta(V)$ can be thought of as a measure of complexity for the geometry of the variety V . The quantity $\Delta(V)$ is a very coarse complexity measure. A recent line of work regarding *algebraic natural proofs* [7, 9] suggests to study the arithmetic circuit complexity of equations for varieties V that correspond to polynomials with small circuit complexity. Having $\Delta(V)$ growing like a polynomial in n is a necessary (but not a sufficient) condition for a variety V to have an algebraic natural proof for non-containment.

1.2 Rigid matrices

A matrix M is (r, s) -rigid if M cannot be written as a sum $R + S$ where $\text{rank}(R) \leq r$ and S contains at most s non-zero entries. Valiant [18] proved that if A is $(\varepsilon n, n^{1+\delta})$ -rigid for some constants $\varepsilon, \delta > 0$ then A cannot be computed by arithmetic circuits of size $O(n)$ and depth $O(\log n)$, and posed the problem of *explicitly* constructing rigid matrices with these parameters, which is still open. It is easy to prove that most matrices have much stronger rigidity parameters: over algebraically closed fields a generic matrix is $(r, (n-r)^2)$ -rigid for any target rank r .

Let \mathbb{F} be an algebraically closed field. Let $A_{r,s} \subseteq \mathbb{F}^{n \times n}$ denote the set of matrices which are not (r, s) -rigid. Let $V_{r,s} = \overline{A_{r,s}}$ denote the Zariski closure of $A_{r,s}$. A geometric study of $V_{r,s}$ was initiated by Kumar, Lokam, Patankar and Sarma [11]. Among other results, they prove that for every $s < (n-r)^2$, $\Delta(V_{r,s}) \leq n^{4n^2}$. A slightly improved (but still exponential) upper bound was obtained by Gesmundo, Hauenstein, Ikenmeyer and Landsberg [8], who also conjectured that for some $\varepsilon, \delta > 0$, $\Delta(V_{\varepsilon n, n^{1+\delta}})$ grows like a polynomial function in n . The following theorem which we prove in this note confirms this conjecture.

► **Theorem 1.** *Let $\varepsilon < 1/25$, and let \mathbb{F} be a field of size at least n^2 . For every large enough n , there exists a non-zero polynomial $Q \in \mathbb{F}[x_{1,1}, \dots, x_{n,n}]$, of degree at most n^3 , which is a non-trivial equation for matrices which are not $(\varepsilon n, \varepsilon n^2)$ -rigid. That is, for every such matrix M , $Q(M) = 0$.*

In fact, the conjecture of [8] was slightly weaker: they conjectured that $\Delta(U)$ is polynomial in n for every irreducible component U of $V_{\varepsilon n, n^{1+\delta}}$. As shown by [11], the irreducible components are in one-to-one correspondence with subsets of $[n] \times [n]$ of size $n^{1+\delta}$ corresponding to possible supports of the sparse matrix S .

As we observe in Remark 6, it is somewhat simpler to show that each of these irreducible components has an equation with a polynomial degree bound. However, since the number of such irreducible components is exponentially large, it is not clear if there is a single equation for the whole variety which is of polynomially bounded degree. We do manage to reverse the order of quantifiers and prove such an upper bound in Theorem 1. This suggests that the set of non-rigid matrices is much less complex than what one may suspect given the results of [11, 8].

1.3 Circuits for linear transformations

The original motivation for defining rigidity was in the context of proving lower bounds for algebraic circuits [18]. If $A \in \mathbb{F}^{n \times n}$ is an $(\varepsilon n, n^{1+\delta})$ -rigid matrix, for any $\varepsilon, \delta > 0$, then the linear transformation represented by A cannot be computed by an algebraic circuit of depth $O(\log n)$ and size $O(n)$.

Every algebraic circuit computing a linear transformation is without loss of generality a *linear* circuit. A linear circuit is a directed acyclic graph that has n inputs labeled X_1, \dots, X_n and n output nodes. Each edge is labeled by a scalar $\alpha \in \mathbb{F}$. Each node computes a linear

function in X_1, \dots, X_n defined inductively. An internal node u with children, v_1, \dots, v_k , connected to it by edges labeled $\alpha_1, \dots, \alpha_k$, computes the linear function $\sum_i \alpha_i \ell_{v_i}$, where ℓ_{v_i} is the linear function computed by v_i , $1 \leq i \leq k$. The size of the circuit is the number of edges in the circuit.

It is possible to use similar techniques to those used in the proof of Theorem 1 in order to prove a polynomial upper bound on an equation for a variety containing all matrices $A \in \mathbb{F}^{n \times n}$ whose corresponding linear transformation can be computed by an algebraic circuit of size at most $n^2/200$ (even without restriction on the depth). Note that this is nearly optimal as any such linear transformation can be computed by a circuit of size n^2 . More formally, we show the following.

► **Theorem 2.** *Let \mathbb{F} be a field of size at least n^2 . For every large enough n , there exists a non-zero polynomial $Q \in \mathbb{F}[x_{1,1}, \dots, x_{n,n}]$, of degree at most n^3 , which is a non-trivial equation for matrices which are computed by algebraic circuit of size at most $n^2/200$.*

Our proofs are based on a dimension counting arguments, and are therefore non-constructive and do not give explicit equations for the relevant varieties. It thus remains a very interesting open problem to provide explicit low-degree equations for any of the varieties considered in this paper. Here “explicit” means a polynomial which has arithmetic circuits of size $\text{poly}(n)$.³ The question of whether such equations exists has a win-win flavor: if they do, this can aid in explicit constructions of rigid matrices, and on the other hand, if all equations are hard, we have identified a family of polynomials which requires super-polynomial arithmetic circuits. Assuming the existence of a polynomial time algorithm for polynomial identity testing, we are able to make this connection formal.

Let PIT denote the set of strings which describe arithmetic circuits (say, over \mathbb{C}) which compute the zero polynomial. It is well known that $\text{PIT} \in \text{coRP}$. Kabanets and Impagliazzo [10] proved that certain circuit lower bounds follow from the assumption that $\text{PIT} \in \text{P}$. As a corollary to Theorem 2, we are able to prove theorem of a similar kind.

► **Corollary 3.** *Suppose $\text{PIT} \in \text{P}$. Then at least one of the following is true:*

1. *There exists a family of n -variate polynomials of degree $\text{poly}(n)$ over \mathbb{C} , which can be computed (as its list of coefficients, given the input 1^n) in PSPACE, which does not have polynomial size constant free arithmetic circuits.*
2. *there exists a family of matrices, constructible in polynomial time with an NP oracle (given the input 1^n), which requires linear circuits of size $\Omega(n^2)$.*

A *constant free arithmetic circuit* is an arithmetic circuit which is only allowed to use the constants $\{0, \pm 1\}$.

A different way to interpret Corollary 3 is as saying that at least one of the following three lower bound results hold: either $\text{PIT} \notin \text{P}$, or (at least) one of the two circuit lower bounds stated in the corollary. We emphasize that the result holds under *any* (even so-called *white box*) derandomization of PIT.

Our statement is similar to, but incomparable with the result of Kabanets and Impagliazzo [10] who proved that if $\text{PIT} \in \text{P}$ then either the permanent does not have polynomial size constant free arithmetic circuits, or $\text{NEXP} \not\subseteq \text{P}/\text{poly}$.

Since $(\varepsilon n, \varepsilon n^2)$ -rigid matrices have linear circuit of size $3\varepsilon n^2$, the last item of Corollary 3 in particular implies a conditional construction of $(\Omega(n), \Omega(n^2))$ -rigid matrices (it is also possible to directly use Theorem 1 instead of Theorem 2 to deduce this result). Unconditional

³ Although one may consider other, informal notions of explicitness which could nevertheless be helpful.

constructions of rigid matrices in polynomial time with an NP oracle were recently given in [2, 3]. However, the rigidity parameters in these papers are not enough to imply circuit lower bounds (furthermore, even optimal rigidity parameters are not enough to imply $\Omega(n^2)$ lower bounds for general linear circuits).

Since it is widely believed that $\text{PIT} \in \text{P}$, the answer to which of the last two items of Corollary 3 holds boils down to the question of whether there exists an equation for non-rigid matrices of degree $\text{poly}(n)$ and circuit size $\text{poly}(n)$. If determining if a matrix is rigid is coNP -hard (as is known for some restricted ranges of parameters [13]), it is tempting to also believe that the equations should not be easily computable, as they provide “proof” for rigidity which can be verified in randomized polynomial time. However, it could still be the case that those equations that have polynomial size circuits only prove the rigidity of “easy” instances.

1.4 Some basic notions in algebraic geometry

For completeness, in this section we define some basic notions in algebraic geometry. A reader who is familiar with this topic may skip to the next section.

Let \mathbb{F} be an algebraically closed field. A set $V \subseteq \mathbb{F}^n$ is called an *affine variety* if there exist polynomials $f_1, \dots, f_t \in \mathbb{F}[x_1, \dots, x_n]$ such that $V = \{\mathbf{x} : f_1(\mathbf{x}) = f_2(\mathbf{x}) = \dots = f_t(\mathbf{x}) = 0\}$. For convenience, in this paper we often refer to affine varieties simply as varieties.

For each variety V there is a corresponding ideal $\mathbf{I}(V) \subseteq \mathbb{F}[x_1, \dots, x_n]$ which is defined as

$$\mathbf{I}(V) := \{f \in \mathbb{F}[x_1, \dots, x_n] : f(\mathbf{x}) = 0 \text{ for all } \mathbf{x} \in V\}.$$

Conversely, for an ideal $I \subseteq \mathbb{F}[x_1, \dots, x_n]$ we may define the variety

$$\mathbf{V}(I) = \{\mathbf{x} : f(\mathbf{x}) = 0 \text{ for all } f \in I\}.$$

Given a set $A \subseteq \mathbb{F}^n$ we may similarly define the ideal $\mathbf{I}(A)$. The (Zariski) *closure* of a set A , denoted \overline{A} , is the set $\mathbf{V}(\mathbf{I}(A))$. In words, the closure of A is the set of common zeros of all the polynomials that vanish on A . It is also the smallest variety with respect to inclusion which contains A . By construction, \overline{A} is a variety, and a polynomial which vanishes everywhere on A is also vanishes on \overline{A} .

Over \mathbb{C} , it is instructive to think of the Zariski closure of A as the usual Euclidean closure. In fact, for the various sets A we consider in this paper (which correspond to sets of “low complexity” objects, e.g., non-rigid matrices or matrices which can be computed with a small circuit), it can be shown that these two notions of closure coincide (see, e.g., Section 4.2 of [4]).

A variety V is called *irreducible* if it cannot be written as a union $V = V_1 \cup V_2$ of varieties V_1, V_2 that are properly contained in V . Every variety can be uniquely written as a union $V = V_1 \cup V_2 \cup \dots \cup V_m$ of irreducible varieties. The varieties V_1, \dots, V_m are then called the *irreducible components* of V .

2 Degree Upper Bound for Non-Rigid Matrices

In this section, we prove Theorem 1. A key component of the proof is the use of the following construction, due to Shpilka and Volkovich, which provides an explicit low-degree polynomial map on a small number of variables, which contains all sparse matrices in its image. For completeness, we provide the construction and prove its basic property.

► **Lemma 4** ([17]). *Let \mathbb{F} be a field such that $|\mathbb{F}| > n$. Then for all $k \in \mathbb{N}$, there exists an explicit polynomial map $\text{SV}_{n,k}(\mathbf{x}, \mathbf{y}) : \mathbb{F}^{2k} \rightarrow \mathbb{F}^n$ of degree at most n such that for any subset $T = \{i_1, \dots, i_k\} \subseteq [n]$ of size k , there exists a setting $\mathbf{y} = \boldsymbol{\alpha}$ such that $\text{SV}(\mathbf{x}, \boldsymbol{\alpha})$ is identically zero on every coordinate $j \notin T$, and equals x_j in coordinate i_j for all $j \in [k]$.*

Proof. Arbitrarily pick distinct $\alpha_1, \dots, \alpha_n \in \mathbb{F}$, and let u_1, \dots, u_n be their corresponding Lagrange's interpolation polynomials, i.e., polynomials of degree at most $n - 1$ such that $u_i(\alpha_j) = 1$ if $j = i$ and 0 otherwise (more explicitly, $u_i(z) = \frac{\prod_{j \neq i} (z - \alpha_j)}{\prod_{j \neq i} (\alpha_i - \alpha_j)}$).

Let $P_i(x_1, \dots, x_k, y_1, \dots, y_k) = \sum_{j=1}^k u_i(y_j) \cdot x_j$, and finally let

$$\text{SV}_{n,k}(\mathbf{x}, \mathbf{y}) = (P_1(\mathbf{x}, \mathbf{y}), \dots, P_n(\mathbf{x}, \mathbf{y})).$$

It readily follows that given $T = \{i_1, \dots, i_k\}$ as in the statement of the lemma, we can set $y_j = \alpha_{i_j}$ for $j \in [k]$ to derive the desired conclusion. The upper bound on the degree follows by inspection. ◀

As a step toward the proof of Theorem 1, we show there is a polynomial map on much fewer than n^2 variables with degree polynomially bounded in n such that its image contains every non-rigid matrix. In the next step, we show that the image of every such polynomial map has an equation of degree $\text{poly}(n)$.

► **Lemma 5.** *There exists an explicit polynomial map $P : \mathbb{F}^{4\epsilon n^2} \rightarrow \mathbb{F}^{n \times n}$, of degree at most n^2 , such that every matrix M which is not $(\epsilon n, \epsilon n^2)$ rigid lies in its image.*

Proof. Let $k = \epsilon n^2$ and let $\mathbf{u}, \mathbf{v}, \mathbf{x}, \mathbf{y}$ denote disjoint tuples of k variables each.

Let U be a symbolic $n \times \epsilon n$ matrix whose entries are labeled by the variables \mathbf{u} , and similarly let V be a symbolic $\epsilon n \times n$ matrix labeled by \mathbf{v} . Let $UV(\mathbf{u}, \mathbf{v}) : \mathbb{F}^{2k} \rightarrow \mathbb{F}^{n \times n}$ be the degree 2 polynomial map defined by the matrix multiplication UV .

Finally, let $P : \mathbb{F}^{4k} \rightarrow \mathbb{F}^{n \times n}$ be defined as

$$P(\mathbf{u}, \mathbf{v}, \mathbf{x}, \mathbf{y}) = UV(\mathbf{u}, \mathbf{v}) + \text{SV}_{n^2,k}(\mathbf{x}, \mathbf{y}),$$

where $\text{SV}_{n^2,k}$ is as defined in Lemma 4.

Suppose now M is a non-rigid matrix, i.e., $M = R + S$ for R of rank ϵn and S which is ϵn^2 -sparse. Decompose $R = U_0 V_0$ for $n \times \epsilon n$ matrix U_0 and $\epsilon n \times n$ matrix V_0 . Let T denote the support of S . For convenience we may assume $|T| = k$ (otherwise, pad with zeros arbitrarily). Let $\boldsymbol{\alpha} \in \mathbb{F}^k$ denote the setting for \mathbf{y} in $\text{SV}_{n^2,k}$ which maps x_1, \dots, x_k to T , and let $\mathbf{s} = (s_1, \dots, s_k)$ denote the non-zero entries of S . Then

$$P(U_0, V_0, \mathbf{s}, \boldsymbol{\alpha}) = U_0 V_0 + S = R + S = M. \quad \blacktriangleleft$$

To complete the proof of Theorem 1, we now argue that the image of any polynomial map with parameters as in Lemma 5 has an equation of degree at most n^3 .

Proof of Theorem 1. Let V_1 denote the subspace of polynomials over \mathbb{F} in n^2 variables of degree at most n^3 . Let V_2 denote the subspace of polynomials over \mathbb{F} in $4\epsilon n^2$ variables of degree at most n^5 . Let P be as in Lemma 5, and consider the linear transformation $T : V_1 \rightarrow V_2$ given by $Q \mapsto Q \circ P$, where $Q \circ P$ denotes the composition of the polynomial Q with the map P , i.e., $(Q \circ P)(\mathbf{x}) = Q(P(\mathbf{x}))$ (indeed, observe that since $\deg(Q) \leq n^3$ and $\deg(P) \leq n^2$, it follows that $\deg(Q \circ P) \leq n^5$).

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We have that $\dim(V_1) = \binom{n^3+n^2}{n^2} \geq n^{n^2}$, whereas $\dim(V_2) = \binom{4\epsilon n^2+n^5}{4\epsilon n^2} \leq (2n^5)^{4\epsilon n^2} < \dim(V_1)$ by the choice of ϵ , so that there exists a non-zero polynomial in the kernel of T , that is, $0 \neq Q_0 \in V_1$ such that $Q_0 \circ P \equiv 0$.

It remains to be shown that for any non-rigid matrix M , $Q_0(M) = 0$. Indeed, let M be a non-rigid matrix. By Lemma 5, there exist $\beta \in \mathbb{F}^{4\epsilon n^2}$ such that $P(\beta) = M$. Thus, $Q_0(M) = Q_0(P(\beta)) = Q_0 \circ P(\beta) = 0$, as $Q_0 \circ P \equiv 0$. ◀

► **Remark 6.** If the support of the sparse matrix is fixed a-priori to some set $S \subseteq [n] \times [n]$ of cardinality at most ϵn^2 , then it is easier to come up with a universal map \tilde{P} from $\mathbb{F}^{3\epsilon n^2} \mapsto \mathbb{F}^{n \times n}$ such that every matrix M whose rank can be reduced to at most ϵn by changing entries in the set S is contained in the image of \tilde{P} . Just consider $\tilde{P}(\mathbf{w}, \mathbf{x}, \mathbf{y}) = UV(\mathbf{u}, \mathbf{v}) + W$, where W is a matrix such that for all $(i, j) \in [n] \times [n]$, if $(i, j) \in S$, then $W(i, j) = w_{i,j}$ and $W(i, j)$ is zero otherwise. Here, each $w_{i,j}$ is a distinct formal variable. Combined with the dimension comparison argument we used in the proof of Theorem 1, it can be seen that there is a non-zero low degree polynomial \tilde{Q} such that $\tilde{Q} \circ \tilde{P} \equiv 0$. This argument provides a (different) equation of polynomial degree for each irreducible component of the variety of non-rigid matrices.

► **Remark 7.** It is possible to use the equation given in Theorem 1, and using the methods of [11], to construct “semi-explicit” $(\epsilon n, \epsilon n^2)$ -rigid matrices. These are matrices whose entries are algebraic numbers (over \mathbb{Q}) with short description, which are non-explicit from the computational complexity point of view. However, such constructions are also known using different methods (see Section 2.4 of [12]).

3 Degree Upper bound for Other Models

The proof of Theorem 2 is omitted from this version and appears in the full version of the paper. The strategy, as before, is to observe that all matrices with a small circuit lie in the image of a polynomial map P on a small number of variables and small degree. Circuits of size s can have many different topologies and thus we first construct a “universal” linear circuit, of size $s' \leq s^4$, that contains as subcircuits all linear circuits of size s . Importantly, s' will affect the degree of P but not its number of variables. We note that this construction of universal circuits is slightly different from similar constructions in earlier work, e.g., in [16]; the key difference being that a naive use of ideas in [16] to obtain the map P seems to incur an asymptotic increase in the number of variables of P , which is unacceptable in our current setting.

Analogous to the proof of Theorem 1, we then observe via a dimension counting argument that the image of the polynomial map P has a equation of degree at most n^3 . This completes the proof of Theorem 2.

Another algebraic object which is closely related to proving circuit lower bounds is the set of three dimensional tensors of high rank. A three dimensional tensor of rank at least r implies a lower bound of r on an arithmetic circuit computing the bi-linear function associated with the tensor. Our arguments also provide polynomial degree upper bounds for the set of tensors of (border) rank at most $n^2/300$. We again omit the details, which appear in the full version of the paper.

A similar method can be used to prove the existence of an equation of degree $\text{poly}(n)$ for three dimensional tensors of *slice rank* (see, e.g., [5]) at most, say, $n/1000$. The existence of such an equations was proved (using different techniques) in [5].

4 Applications to Circuit Lower Bounds

In this section we prove Corollary 3. The strategy of the proof is simple: the proof of Theorem 2 implies a PSPACE algorithm which produces a sequence of polynomials which are equations for the set of matrices with small linear circuits. If those equations require large circuits, we are done, and if not, then there exists an equation with small circuits which (assuming $\text{PIT} \in \text{P}$) can be found using an NP-oracle. Using, once again, the assumption that $\text{PIT} \in \text{P}$, we can also find deterministically a matrix on which the equation evaluates to non-zero, which implies the matrix requires large linear circuits.

There are some technical difficulties involved with this plan which we now describe. The first problem is that even arithmetic circuits of small size can have large description as bit strings, due to the field constants appearing in the circuits. To prevent this issue, we only consider *constant free* arithmetic circuits, which are only allowed inputs labeled by $\{0, \pm 1\}$ (but can still compute other constants in the circuit using arithmetic operations).

The second problem is that, in order to be able to find a non-zero of the equation in the last step of the algorithm (using the mere assumption that $\text{PIT} \in \text{P}$), we need not only the size of the circuit but also its *degree* to be bounded by $\text{poly}(n)$. Of course, by Theorem 2 there exists such a circuit, but we need to be able to prevent a malicious prover from providing us with a $\text{poly}(n)$ size circuit of exponential degree, and it is not known how to compute the degree of a circuit in deterministic polynomial time, even assuming $\text{PIT} \in \text{P}$. To solve this issue, we use an idea of Malod and Portier [14], who showed that any polynomial with circuit of size $\text{poly}(n)$ and degree d also has a *multiplicatively disjoint* (MD) circuit of size $\text{poly}(n, d)$. An MD circuit is a circuit in which any multiplication gate multiplies two disjoint subcircuits. This is a syntactic notion which is easy to verify efficiently and deterministically, and an MD circuit of size s is guaranteed to compute a polynomial of degree at most s .

A final technical issue is that the notion of MD circuits does not fit perfectly within the framework of constant free circuits. Therefore we use the notion of “almost MD” circuits, which allow for the case which the inputs to a multiplication gates are not disjoint, as long as at least one of them is the root of a subcircuit in which only constants appear.

► **Definition 8.** *We say a gate v in a circuit is constant producing (CP) if in the subcircuit rooted at v , all input nodes are field constants.*

An almost-MD circuit is a circuit where every multiplication gate either multiplies two disjoint subcircuits, or at least one of its children is constant producing.

► **Lemma 9.** *Suppose f is an n -variate polynomial of degree $\text{poly}(n)$ which has a constant free arithmetic circuit of degree $\text{poly}(n)$. Then f has a constant free almost-MD circuit of size $\text{poly}(n)$.*

The proof of Lemma 9 appears in the full version of the paper.

For circuits which compute low-degree polynomials, the mere existence of an algorithm for the decision version of PIT allows one to construct an algorithm for the search version. The following lemma is standard, and its proof appears in the full version of the paper.

► **Lemma 10.** *Suppose $\text{PIT} \in \text{P}$. Then there is a polynomial time algorithm that given a non-zero almost-MD arithmetic circuit C of size s computing an n -variate polynomial, finds in time $\text{poly}(n, s)$ an element $\mathbf{a} \in \mathbb{C}^n$ such that $C(\mathbf{a}) \neq 0$.*

As we noted above, the assumption that C is almost-MD was used in Lemma 10 to bound the degree of the circuit. It is also useful because it is easy to decide in deterministic polynomial time whether a circuit is almost-MD. We now complete the proof of Corollary 3.

Proof of Corollary 3. For every n , the proof of Theorem 2 provides an equation Q_n for the set of $n \times n$ matrices with small linear circuits. This polynomial can be found by solving a linear system of equations in a linear space whose dimension is $\exp(\text{poly}(n))$. Using standard, small space algorithm for linear algebra [6, 1], this implies that there exists a fixed PSPACE algorithm which, on input 1^n , outputs the list of coefficients of the polynomial Q_n .

Consider now the family $\{Q_n\}_{n \in \mathbb{N}}$. If for any constant $k \in \mathbb{N}$ there exist infinitely many $n \in \mathbb{N}$ such that Q_n requires circuits of size at least n^k , it follows (by definition) that the PSPACE algorithm above outputs a family of polynomials with super-polynomial constant-free arithmetic circuits.

We are thus left to consider the case that there exists a constant $k \in \mathbb{N}$ such that for all large enough $n \in \mathbb{N}$, Q_n can be computed by circuits of size n^k . By Lemma 9, we may assume without loss of generality that these circuits are almost-MD circuits. Further suppose $\text{PIT} \in \text{P}$. We will show how to construct a matrix in polynomial time with an NP oracle which requires large linear circuits.

Consider the language L of pairs $(1^n, x)$ such that there exists a string y of length at most n^k such that xy describes an almost-MD circuit C such that C is non-zero, and $C \circ U \equiv 0$, where U is the polynomial map given in the proof of Theorem 2.

Assuming $\text{PIT} \in \text{P}$, the language L is in NP, and by assumption for every large enough n there exists such a circuit. Thus, we can use the NP oracle to construct such a circuit C bit by bit. Finally, using Lemma 10 we can output a matrix M such that $C(M) \neq 0$.

By the properties of the circuit C and the map U , M does not have linear circuits of size less than $n^2/200$. ◀

Many variations of Corollary 3 can be proved as well, with virtually the same proof. By slightly modifying the language L used in the proof, it is possible to prove the same result even under the assumption $\text{PIT} \in \text{NP}$ (recall that $\text{PIT} \in \text{coRP}$). A similar statements also holds over finite fields of size $\text{poly}(n)$, in which case the proof is simpler since there are no issues related to the bit complexity of the first constants. Finally, an analog of Corollary 3 also holds for tensor rank: that is, assuming $\text{PIT} \in \text{P}$, either there exists a construction of a hard polynomial in PSPACE, or an efficient construction with an NP oracle of a 3-dimensional tensor of rank $\Omega(n^2)$. We remark that for tensors of large rank there are no analogs of [2, 3], i.e., there do not exist even constructions with an NP oracle of tensors with slightly super-linear rank.

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