On Identity Testing and Noncommutative Rank Computation over the Free Skew Field

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Abstract

The identity testing of rational formulas (RIT) in the free skew field efficiently reduces to computing the rank of a matrix whose entries are linear polynomials in noncommuting variables \[23\]. This rank computation problem has deterministic polynomial-time white-box algorithms \[20, 25\] and a randomized polynomial-time algorithm in the black-box setting \[14\]. In this paper, we propose a new approach for efficient derandomization of black-box RIT. Additionally, we obtain results for matrix rank computation over the free skew field and construct efficient linear pencil representations for a new class of rational expressions. More precisely, we show:

- Under the hardness assumption that the ABP (algebraic branching program) complexity of every polynomial identity for the \(k \times k\) matrix algebra is \(2^{\Omega(k)}\) \[9\], we obtain a subexponential-time black-box RIT algorithm for rational formulas of inversion height almost logarithmic in the size of the formula. This can be seen as the first “hardness implies derandomization” type theorem for rational formulas.

- We show that the noncommutative rank of any matrix over the free skew field whose entries have small linear pencil representations can be computed in deterministic polynomial time. While an efficient rank computation was known for matrices with noncommutative formulas as entries \[19\], we obtain the first deterministic polynomial-time algorithms for rank computation of matrices whose entries are noncommutative ABPs or rational formulas.

- Motivated by the definition given by Bergman \[7\], we define a new class of rational functions where a rational function of inversion height at most \(h\) is defined as a composition of a noncommutative r-skewed circuit (equivalently an ABP) with inverses of rational functions of this class of inversion height at most \(h - 1\) which are also disjoint. We obtain a polynomial-size linear pencil representation for this class which gives a white-box deterministic polynomial-time identity testing algorithm for the class.

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1 Introduction

In algebraic circuit complexity the basic arithmetic operations are additions, multiplications, and inverses. Using these arithmetic operations algebraic circuits compute either polynomials or rational functions. A subarea of algebraic complexity is noncommutative computation where the multiplication of variables is not commutative and the set of monomials (over the variables) form a free monoid. Noncommutative circuits/formulas with only addition and multiplication gates compute noncommutative polynomials in the free algebra.

In the commutative case, inverses are well understood but in noncommutative computation inverses are quite subtle. To elaborate, it is known that any commutative rational circuit can be efficiently transformed into the form \( fg^{-1} \) where \( f \) and \( g \) are polynomials that are computed by circuits [32]. However, noncommutative rational expressions such as \( x^{-1} + y^{-1} \) or \( xy^{-1}x \) cannot be represented as \( fg^{-1} \) or \( f^{-1}g \). If we have nested inverses then it makes the rational expression more complicated, for example \( (z + xy^{-1}x)^{-1} - z^{-1} \). Moreover, a noncommutative rational expression is not always defined on a matrix substitution. For a noncommutative rational expression \( \tau \), its domain of definition \( \text{dom}(\tau) \) is the set of all matrix tuples (of any dimension) where \( \tau \) is defined. Two rational expressions \( \tau_1 \) and \( \tau_2 \) are equivalent if they agree on \( \text{dom}(\tau_1) \cap \text{dom}(\tau_2) \). This defines an equivalence relation on noncommutative rational expressions (with nonempty domains of definition). Amitsur used this in defining the noncommutative rational functions [1]. This object plays an important role in the study of noncommutative algebra [1, 12], control theory [27], and algebraic automata theory [34].

Computationally, rational functions are represented by noncommutative arithmetic circuits or formulas using addition, multiplication, and inverse gates [23]. The inversion height of a rational formula is the maximum number of inverse gates in a path from an input gate to the output gate. It is known that the inversion height of a rational formula of size \( s \) is bounded by \( O(\log s) \) [23]. Hrubeš and Wigderson introduced the rational identity testing problem (RIT) of testing the equivalence of two rational formulas [23]. It is the same as testing whether a rational formula computes the zero function in the free skew field. In other words, the problem is to decide whether there is a matrix tuple (of some dimension) such that the rational formula evaluates to nonzero on it. Rational expressions exhibit peculiar properties which seem to make the RIT problem quite different from polynomial identity testing. The apparent lack of canonical representations such as the sum of monomials representation for polynomials and the use of nested inverses in noncommutative rational expressions complicate it. For example, the rational expression \( (x + xy^{-1}x)^{-1} + (x + y)^{-1} - x^{-1} \) of inversion height two is a rational identity, known as Hua’s identity [24].

A second characterization of the free skew field elements is due to Cohn [12]. A linear pencil \( L \) of size \( s \) over noncommuting variables \( x = \{x_1, \ldots, x_n\} \) is an \( s \times s \) matrix whose entries are linear forms in \( x \) variables, i.e. \( L = A_0 + \sum_{i=1}^s A_i x_i \), where each \( A_i \) is an \( s \times s \) matrix over the field \( \mathbb{F} \). Cohn has shown that for every \( \tau \) in \( \mathbb{F}[x] \), there is a linear pencil \( L \) such that \( \tau \) is an entry of the matrix inverse of \( L \). More generally, \( \tau \) has a linear pencil representation of size \( s \), if for vectors \( c, b \in \mathbb{F}^s \) and an \( s \times s \) linear pencil \( L \), \( \tau = c^t L^{-1} b \) where \( c^t \) is the transpose of \( c \). Hrubeš and Wigderson give an efficient reduction from RIT to the singularity testing problem of linear pencils [23]. In particular, if \( \tau \) is a rational formula of size \( s \), they showed that \( \tau \) has a linear pencil representation \( L \) of size at most \( 2s \) such that \( \tau \) is defined on a matrix tuple if and only if \( L \) is invertible on that tuple [23]. Using this connection, they reduce the RIT problem to the problem of testing whether a given linear
pencil is invertible over the free skew field in deterministic polynomial time. The latter is the noncommutative Singular problem, whose commutative analog is the symbolic determinant identity testing problem. The deterministic complexity of symbolic determinant identity testing is completely open in the commutative setting [26]. In contrast, the Singular problem in noncommutative setting has deterministic polynomial-time algorithms in the white-box model due to the works of Garg et al. [20] which is based on operator scaling and that of Ivanyos et al. [25] which is based on the second Wong sequence and a constructive version of regularity lemma. As a consequence, a deterministic polynomial-time white-box RIT algorithm follows.

A central open problem in this area is to design an efficient deterministic algorithm for noncommutative Singular problem in the black-box case [20]. The algorithms by Garg et al. [20] and Ivanyos et al. [25] are inherently sequential and we believe that they are unlikely to be helpful for black-box algorithm design. It is well-known [20] that an efficient black-box algorithm (via a hitting set construction) for Singular would generalize the celebrated quasi-NC algorithm for bipartite matching significantly [17]. There is a randomized polynomial-time black-box algorithm for this problem [14].

Even for the RIT problem (which could be easier than the noncommutative Singular problem), there is limited progress towards designing efficient deterministic black-box algorithm. In fact, only very recently a deterministic quasipolynomial-time black-box algorithm for identity testing of rational formulas of inversion height two has been designed [3]. It is interesting to note that in the literature of identity testing, the noncommutative Singular problem and the RIT problem stand among rare examples where deterministic polynomial-time white-box algorithms are designed but for the black-box case no deterministic subexponential-time algorithm is known.

\[\text{Remark 1.}\] For noncommutative polynomials computed by polynomial-size arithmetic circuits, efficient randomized polynomial identity testing algorithms are known either for polynomial degree bound or for exponential sparsity bound [9, 5]. In contrast, the complexity of testing the identity of rational circuits is completely open. In fact, even in the white-box setting we do not have a randomized subexponential-time algorithm.

### 1.1 Derandomization of RIT from the hardness of polynomial identities

In this paper, we propose a new approach to the RIT problem in the black-box case under a suitable hardness assumption, based on the following conjecture due to Bogdanov and Wee [9, Section 6.2].

\[\text{Conjecture 2.}\] The ABP complexity (i.e. the minimum size of an algebraic branching program) of a polynomial identity for the $k \times k$ matrix algebra $\mathcal{M}_k(\mathbb{F})$ is $2^{\Omega(k)}$.

The conjecture implies that ABPs of size $s$ cannot evaluate to zero on all $O(\log s)$-dimensional matrices. Bogdanov and Wee [9] also observed that if the conjecture holds then there is an $s^{O(\log^2 s)}$-time black-box PIT for noncommutative ABPs\(^1\) and as supportive evidence showed that the conjecture is indeed true for normal identities (of which the standard identity is a special case), and the identity of algebraicity.

Consider the following variant of the usual hitting set definition.

\(^1\) Independent of the conjecture, Forbes-Shpilka [18] obtained an $s^{O(\log s)}$-time black-box PIT for noncommutative ABPs.
Definition 3. For a class of rational formulas $\mathcal{R}$, we say that a hitting set $H$ is strong if for any nonzero formula $r \in \mathcal{R}$, there exists a matrix tuple $p \in H$ such that $r(p)$ is invertible.

In the following theorem, we show an efficient derandomization of RIT assuming Conjecture 2. This can be seen as the first "hardness implies derandomization" type result for rational formulas.

Theorem 4. If Conjecture 2 is true then we can construct a strong hitting set of size $(s \cdot h(\log s)^{2h+2})O(h(\log s)^{2h+2})$ for rational formulas $r$ of size $s$ over $n$ variables and inversion height $h$ in deterministic $(s \cdot h(\log s)^{2h+2})O(h(\log s)^{2h+2})$ time for some constant $\gamma > 1$. This result holds over infinite or sufficiently large finite fields and $h \leq \beta(\log s/\log \log s)$ for any $0 < \beta < 1$.

As a special case for $h = O(1)$, this gives a quasipolynomial-size hitting set. To get a subexponential-size bound $2^s$ on the hitting set, for $\delta$ in $(0, 1)$, we can allow $h \leq c_6(\log s/\log \log s)$. Here $c_6 \in (0, 1)$ is a constant that depends on $\delta$.

As already mentioned, the inversion height of size $s$ rational formula is bounded by $O(\log s)$ [23]. Therefore, Theorem 4 solves the RIT problem in an almost general setting.

We believe that the main interesting point about Theorem 4 is that it relates the black-box RIT derandomization that involves handling nested inverses with a problem purely for noncommutative polynomials: we can obtain a deterministic black-box RIT algorithm by showing an exponential size lower bound for ABPs computing any polynomial identity for matrix algebras. Over the years such hardness assumptions have proved to be useful in designing deterministic algorithms for problems related to identity testing [21, 26, 6, 15, 11, 29].

Proof Sketch

The first step in proving Theorem 4 is a variable reduction step that shows the identity testing of a rational formula $r$ of inversion height $h$ can be reduced to the identity testing of another rational formula $r'$ over $(2h + 1)$ variables in a black-box manner. Notice that for noncommutative polynomials (for which $h = 0$), such a reduction is standard and given by $x_i \rightarrow y_0 y_i y_0$ where $y_0, y_1$ are new noncommutative variables. We prove it by induction on $h$. In fact, we use a stronger inductive hypothesis that roughly says that for every nonzero rational formula $r$ of inversion height $h$, there also exists a $(2h + 1)$-tuple of matrices $(q_{00}, \ldots, q_{0h}, q_{10}, \ldots, q_{1h})$ such that $r(p_{1}, \ldots, p_{n})$ is invertible and for each $i \in [n]$, $p_i = \sum_{j=0}^{h} q_{ij} q_{j0}$. Once we assume the inductive hypothesis for inversion height $h - 1$, for each rational formula $r$ of inversion height $h$, we get a matrix tuple of the form $p = (p_{1}, \ldots, p_{n})$ where $p_i = \sum_{j=0}^{h-1} q_{ij} q_{j0}$ such that $r$ is defined on $p$. Then, using concepts from matrix coefficient realization theory we construct the nonzero generalized series $\tau(x + p)$ [34]. Now, we can use the standard bivariate encoding trick on $\tau(x + p)$ to complete the variable-reduction step.

The next important step that we establish is that if Conjecture 2 is true then for any rational formula $r$ of size $s$ and inversion height $h$, one can find a matrix tuple $p$ of dimension $(\gamma \log s)^{h+1}$ (for some constant $\gamma$) such that $r(p)$ is an invertible matrix. This is done via induction on $h$ and a bootstrapping argument. For the base case, we take $h = 0$. In this case the rational formula is also an ABP of size $s$ and Conjecture 2 confirms that $r$ is nonzero on a generic matrix tuple $p$ of dimension $O(\log s)$. Also $r(p)$ is invertible by an application of Amitsur’s theorem [1]. Inductively we assume that we can find such a matrix tuple $q$ of dimension $d_{h-1} \leq (\gamma \log s)^h$ for any rational formula $r$ of inversion height at most $h - 1$. 

In the following theorem, we show an efficient derandomization of RIT assuming Conjecture 2. This can be seen as the first “hardness implies derandomization” type result for rational formulas.
and size at most \( s \). An easy observation shows that given a rational formula \( \tau \) of inversion height \( h \), \( \tau \) is defined on such a matrix tuple \( q \). By matrix coefficient realization theory [34] we can construct the nonzero generalized series \( \tau(x + q) \) by expanding \( \tau \) around the point \( q \). Substituting the variables \( x_1, x_2, \ldots, x_n \) by symbolic generic matrices over noncommuting variables \( Z(1), \ldots, Z(n) \) of dimension \( d_{h-1} \), we observe that each entry of the output matrix \( \tau(Z + q) \) is a recognizable series computed by a small size algebraic automaton. By a standard result in algebraic automata theory, generally attributed to Schützenberger [16, Corollary 8.3, Page 14], we know that this series is nonzero if and only if it is nonzero when truncated to the degree matching the size of the algebraic automaton. By Conjecture 2, we can infer that the truncated series is nonzero on generic matrices of dimension roughly \( \approx \log(sd_{h-1}) \). A simple scaling trick shows that the full (infinite)-series is also nonzero on generic matrices of same dimension. This determines the dimension of the generic matrices on which the rational formula \( \tau \) is nonzero. Moreover the rational formula evaluates to an invertible matrix on generic matrix substitution of that dimension. This is a consequence of Amitsur’s theorem [1].

Once we have these two steps, the rest of the proof is straightforward. Given nonzero \( \tau \) over the variables \( x_1, \ldots, x_n \) of height \( h \), we apply the variable reduction step to construct nonzero \( \tau' \) of height \( h \) (and roughly of same size) over \( 2(h+1) \) variables \( \{y_0, y_1, \ldots, y_{h0}, y_{h1}\} \). Now we apply the second step that says that \( \tau' \) is nonzero (and hence invertible) on generic matrices over \( Z \) variables of dimension \( (\gamma \log s)^{h+1} \). We also make use of the fact that \( \tau'(y) \) has a small-size linear pencil. To construct the final hitting set, we just need to hit two sparse polynomials of sparsity bound roughly \( \left( snh(\gamma \log s)^{2h+2}\right)^{O(h(\gamma \log s)^{2h+2})} \) and this can be done by applying the standard result of sparse polynomial hitting set construction [28].

### 1.2 Noncommutative rank of matrices over the free skew field

For a matrix \( M = (g_{i,j})_{m \times m} \) over the free skew field \( \mathbb{F}\langle x \rangle \), its noncommutative rank (denoted by \( \text{ncrank}(M) \)) is the least positive integer \( r \leq m \) such that \( M = PQ \) for an \( m \times r \) matrix \( P \) and an \( r \times m \) matrix \( Q \) over \( \mathbb{F}\langle x \rangle \). This is also called the inner rank. If \( r = m \), then \( M \) is invertible in \( \mathbb{F}\langle x \rangle \).

Indeed, a fundamental result of Cohn [13] showed that for any matrix \( M = (g_{i,j})_{m \times m} \) over the noncommutative ring \( \mathbb{F}(x) \) such that \( \text{ncrank}(M) = r \), there exists an \( m \times r \) matrix \( P \) and an \( r \times m \) matrix \( Q \) over \( \mathbb{F}(x) \).

As already mentioned, the problem of computing the noncommutative rank of a linear matrix admits deterministic polynomial-time white-box algorithms [20, 25]. If the matrix entries consist of some higher degree terms, one can use Higman’s trick [22] to reduce it to computing rank of a linear matrix. Consider the following well-known example of a \( 2 \times 2 \) matrix [19]:

\[
\begin{pmatrix}
1 & x \\
y & z + xy
\end{pmatrix}.
\]

Higman’s trick reduces it to another \( 3 \times 3 \) linear matrix preserving the complement of the noncommutative rank in the following way:

\[
\begin{pmatrix}
1 & x \\
y & z + xy \\
0 & 0 & 1
\end{pmatrix} \mapsto \begin{pmatrix}
1 & x & 0 \\
y & z + xy & 0 \\
0 & 0 & 1
\end{pmatrix} \mapsto \begin{pmatrix}
1 & x & 0 \\
y & z & x \\
0 & -y & 1
\end{pmatrix}.
\]

However, it would not be efficient in general. In [19, Proposition A.2], the authors showed an effective use of Higman’s trick to efficiently reduce it to the rank computation of a linear matrix when the entries are computed by noncommutative formulas.
In this paper, we consider matrix rank computation over the free skew field in a more general setting. In particular, we obtain an efficient reduction to the rank computation of a linear matrix even when the entries are free skew field elements computed by small linear pencils. More precisely, we show the following.

**Theorem 5.** Let $M = (g_{i,j})_{m \times m}$ be a matrix such that for each $i,j \in [m]$, $g_{i,j}$ in $\mathbb{F}[x_1, \ldots, x_n]$ has a linear pencil of size at most $s$. Then, the noncommutative rank of $M$ can be computed in deterministic $\text{poly}(m,n,s)$ time. Moreover, in deterministic $\text{poly}(m,n,s)$ time, we can output a matrix tuple $\bar{T} = (T_1, \ldots, T_n)$ of dimension $d$ such that the matrix the rank of matrix $M(\bar{T})$ is $d \cdot \text{nccrank}(M)$. The field $\mathbb{F}$ could be infinite or sufficiently large finite field.

We note an earlier result of Garg et al. [19], showing efficient matrix rank computation when the entries are noncommutative polynomials that are computed by formulas. The polynomial-time deterministic algorithm of Theorem 5 is a two-fold strengthening of this: Since noncommutative ABPs have polynomial-size linear pencils [23], the algorithm can compute the rank of matrices with entries that are polynomials computed by ABPs. Moreover, since noncommutative rational formulas also have polynomial-size linear pencils [23], the algorithm computes the rank of matrices whose entries are computed by small rational formulas.

**Proof Sketch**

The basic principle of our proof is to reduce the problem to the rank computation of a linear matrix. However, there is no clear notion of degree reduction for arbitrary elements over the free skew field. This forces us to find a new approach of constructing this linear matrix efficiently that can also handle a matrix of skew field entries as input. The main idea here is to show that the linear pencil representation enjoys the following closure property. Let $A$ be an $m \times m$ generic matrix over $m^2$ indeterminates. Substituting each indeterminate $A_{ij}$ by a free skew field element that has a linear pencil of size at most $s$, suppose we obtain the matrix $M$. We show that there is an efficiently computable linear matrix $L$ such that $\text{nccrank}(L) = m^2 s + \text{nccrank}(M)$. Somewhat surprisingly, the construction of $L$ turns out to be relatively simple and elegant.

There are many equivalent notions of noncommutative rank for linear matrices (for example, see [25, 20]). A notion of particular interest is the blow-up definition that is crucial in the algorithm of Ivanyos et al. [25]. The blow-up notion enables to find a matrix tuple on which the maximum rank is achieved. We extend this notion and introduce a blow-up definition for noncommutative rank (denoted by $\text{nccrank}^*$) of matrices with free skew field entries. We show that for any matrix $M$ of free skew field entries, $\text{nccrank}(M) = \text{nccrank}^*(M)$. Introduction of the blow-up definition allows us to find efficiently the matrix tuple $T$ of dimension $d$ such that the rank of $M(T)$ is $d \cdot \text{nccrank}(M)$. One can view the blow-up definition in this case as an extension of the theory developed by Derksen and Makam [14] for the linear case. This extension could be of independent mathematical interest.

### 1.3 Linear pencil representations for a new class of rational functions

The study of linear pencils seem to be the key in understanding several basic questions in rational function theory [23, 19, 25, 34, 14]. Our main motivation here is to understand the relation between the linear pencil representations of rational functions and their representations using basic arithmetic operations. Let $\text{RF}, \text{LR}, \text{RC}$ denote, respectively, the class of
polynomial-size rational formulas, the class of rational functions that have polynomial-size linear pencil representations, and the class of polynomial-size rational circuits. Hrubeš and Wigderson [23] have shown an exponential size lower bound for rational formulas computing an entry of the inverse of a symbolic matrix. Moreover, they show that each entry of the inverse of a symbolic matrix is computable by a rational circuit of polynomial size. Therefore, the current known relation is \( RF \subseteq LR \subseteq RC \).

Following Bergman [7], a noncommutative rational function \( \tau(x) \) of inversion height at most \( h \) can be inductively defined as \( \tau(x) = f(x_1, \ldots, x_n, g_1^{-1}, \ldots, g_m^{-1}) \), where \( f \) is a noncommutative polynomial and \( g_1, \ldots, g_m \in \mathbb{F}[x] \) are rational functions of inversion height \( \leq h - 1 \). Based on this, we define rational \( r \)-skewed circuits.

**Definition 6.** A rational \( r \)-skewed circuit of inversion height \( 0 \) is a noncommutative \( r \)-skewed circuit\(^2\) which is also a noncommutative ABP. Inductively, we define \( \tau(x) = f(x_1, \ldots, x_n, g_1^{-1}, \ldots, g_m^{-1}) \) as a rational \( r \)-skewed circuit of inversion height at most \( h \) if \( f(x, y_1, \ldots, y_m) \) is a noncommutative \( r \)-skewed circuit \( (m \geq 0) \) and for each \( i \in [m] \), \( g_i(x) \) is a rational \( r \)-skewed circuit of inversion height \( \leq h - 1 \).

Let \( R\text{-}rSC \) denote the class of all rational functions computable by polynomial-size rational \( r \)-skewed circuits. Inspecting the polynomial size rational circuit for symbolic matrix inverse [23], we notice that each entry of the inverse of a polynomial-size symbolic matrix can indeed be computed by a polynomial-size rational \( r \)-skewed circuit. Hence \( LR \subseteq R\text{-}rSC \). What is the exact expressiveness power of the class \( LR \)? In particular, is it true that \( LR = R\text{-}rSC \)? It now suffices to show that \( R\text{-}rSC \subseteq LR \). While we are unable to answer this completely, we show such a containment under additional structural restrictions on the circuit.

**Definition 7.** An inversely disjoint rational \( r \)-skewed circuit of inversion height \( 0 \) is a noncommutative \( r \)-skewed circuit (which is also an ABP). Inductively, we define \( \tau(x) = f(x_1, \ldots, x_n, g_1^{-1}, \ldots, g_m^{-1}) \) as an inversely disjoint rational \( r \)-skewed circuit of inversion height at most \( h \) if \( f(x, y_1, \ldots, y_m) \) is a noncommutative \( r \)-skewed circuit \( (m \geq 0) \) and for each \( i \in [m] \), \( g_i(x) \) is a inversely disjoint rational \( r \)-skewed circuit of inversion height \( \leq h - 1 \) and for all \( i \neq j \), the circuits of \( g_i \) and \( g_j \) are disjoint.

Let \( \text{ID-R-}r\text{SC} \) be the class of rational functions computed by polynomial-size inversely disjoint \( r \)-skewed circuits. This class contains rational formulas, ABPs. We are able to give polynomial-size linear pencil representations for this class.

**Theorem 8.** Over any field, an inversely disjoint rational \( r \)-skewed circuit of size \( s \) has a linear pencil representation of size \( O(s^2) \) which can be computed in deterministic polynomial time from the given circuit.

This gives the following containment:

\[ RF \subseteq \text{ID-R-}r\text{SC} \subseteq LR \subseteq R\text{-}rSC \subseteq RC, \]

where we know at least one of the first two containment is proper. We do not know any unconditional separation between \( RF \) and \( \text{ID-R-}r\text{SC} \). This question is somewhat similar in spirit to the separation of noncommutative formulas and ABPs which is still open [30, 10, 33]. However, a simple inductive argument shows that functions of inversion height \( h \)

\(^2\) Usually in the literature they are called right-skew circuits. In this paper, we refer to them as right-skewed circuits and reserve the word “skew” for skew fields.
in ID-R-rSC can be computed by rational formulas of size \( s^{O(h \log s)} \). By standard techniques, a noncommutative r-skewed circuit of size \( s \) can be computed by a formula of size \( s^{O(\log s)} \). Consider an inversely disjoint r-skewed circuit \( \tau(x, g_1^{-1}, \ldots, g_m^{-1}) \) where each \( g_i \in \text{ID-R-rSC} \) of inversion height \( \leq h - 1 \) for each \( 1 \leq i \leq m \). Inductively, each \( g_i \) has a rational formula of size \( s^{O((h-1) \log s)} \). Therefore, the size of the rational formula computing \( \tau \) can be at most \( s^{O(h \log s)} \). If \( h = O(\log s) \), we then have a quasipolynomial-size formula simulation for this class. However, unlike rational formulas [23], it is not clear whether \( h \) can be taken as \( O(\log s) \) for a general inversely disjoint r-skewed circuit of size \( s \).

Using Theorem 8, the following corollary is obtained by the application of rank computation algorithm in [25]. For the black-box case, we can apply the algorithm in [14]. In the proof of the corollary we also mention how to apply the algorithm in [25] for the black-box case and get an efficient randomized algorithm over finite fields as well.

**Corollary 9.** Let \( \mathbb{F} \) be infinite or a sufficiently large field. For an inversely disjoint rational r-skewed circuit of size at most \( s \) and over \( n \) variables, we can decide whether or not it computes zero in \( \mathbb{F}\langle x \rangle \) in deterministic \( \text{poly}(s,n) \) time in white-box, and in randomized \( \text{poly}(s,n) \) time in black-box.

**Proof Sketch**

The proof is based on a composition lemma that computes an efficient linear pencil for \( f(x, g_1^{-1}, \ldots, g_m^{-1}) \) from the linear pencils of \( f(x, y) \) and \( g_1^{-1}, \ldots, g_m^{-1} \). It turns out that the proof of this composition result is more subtle than the usual proofs of the linear pencil constructions for rational formulas [23, 34].

We first elaborate on the composition lemma. Let \( L \) be an \( s \times s \) linear pencil over \( x_1, \ldots, x_n \) and \( y_1, \ldots, y_m \). Let \( f_{i,j} = (L^{-1})_{i,j} \) for \( i,j \in [s] \). Let \( g_1, \ldots, g_m \) be rational functions over \( x_1, \ldots, x_n \) such that each \( g_i \) has a linear pencil \( L_k \) of size at most \( s' \). Then we can construct a single linear pencil \( \tilde{L} \) of size at most \( ms'+m+2s^2+s \) in \( \text{poly}(s',s,m,n) \)-time such that

\[
(\tilde{L}^{-1})_{2s+\hat{s}+i,2s+\hat{s}+j} = f_{i,j}(x, g_1^{-1}, \ldots, g_m^{-1}) \quad \text{for} \ i,j \in [s], \text{where} \ \hat{s} = ms' + m.
\]

Given a rational function \( \tau \) computed by an inversely disjoint rational r-skewed circuit of size at most \( s \), we consider the rational function \( \tau^{-1} \) which is still in the same class (with inversion height increased by one). Using the composition result, we construct a linear pencil of size \( O(s') \) for \( \tau^{-1} \). Notice that \( \tau(x,y) \) is a polynomial computed by an ABP or a r-skewed circuit and it has a polynomial-size linear pencil [23]. Using a standard idea, \( \tau^{-1} \) also has a small linear pencil \( L \) which we use as the input to the composition lemma along with the inductively constructed linear pencils for \( g_1, \ldots, g_m \).

The final linear pencil \( \tilde{L} \) which is the outcome of the composition lemma has the additional property that for any matrix tuple \( \tau(p), \tau^{-1}(p) \) is defined if and only if \( \tilde{L}(p) \) is invertible. Since \( \tau \neq 0 \) if and only if \( \tau^{-1} \) is defined [1], we can now use the algorithm for noncommutative Singular problem [25] on the linear pencil \( \tilde{L} \) to check the identity of \( \tau \).

**Organization**

In Section 2, we mainly provide brief background on linear pencils and its connection with the rational identity testing problem, and also present some results in matrix coefficient realization theory. We prove Theorem 4 in Section 3. The proof of Theorem 5 is given in Section 4.1. We give the proof of Theorem 8 in Section 5. We state some open questions in Section 6.
2 Preliminaries

2.1 Linear pencils and rational functions

Let $\mathbb{F}$ be a field. A linear pencil $L$ of size $s$ over noncommuting $x = \{x_1, \ldots, x_n\}$ variables is a $s \times s$ matrix where each entry is a linear form in $x$. That is, $L = A_0 + \sum_{i=1}^n A_i x_i$ where each $A_i$ is in $M_s(\mathbb{F})$. Evaluation of a linear pencil at a matrix tuple $p = (p_1, \ldots, p_n)$ in $M_n(\mathbb{F})$ is defined using the Kronecker (tensor) product: $L$ evaluated at $p$ is $A_0 \otimes I_n + \sum_{i=1}^n A_i \otimes p_i$.

Given a linear pencil $L$, the noncommutative SINGULAR problem is to decide whether there is a tuple $p$ in $M_n^m(\mathbb{F})$ of $m \times m$ matrices for some $m$ such that the output matrix $L$ evaluated at $p$ is invertible.

A rational function $r$ in $\mathbb{F}<x>$ has a linear pencil representation $L$ of size $s$ if $r = c^tL^{-1}b$ for vectors $c, b \in \mathbb{F}^n$. Following is the re-statement of Proposition 7.1 proved in [23].

Proposition 10. Let $r$ be a rational function given by a rational formula of size $s$. Then $r$ can be represented $(L^{-1})_{i,j}$ for $i, j \in [s]$ where $L$ is a linear pencil of size at most $2s$. Moreover, $r$ is nonzero if and only if $L$ is invertible.

Clearly in the above proposition the choice for $c, b$ are the indicator vectors $e_i$ and $e_j$.

We also use the following classical result of Amitsur [1] in this paper.

Theorem 11 ([1]). Let $r$ be a rational function which is nonzero on $M_k(\mathbb{F})$ where $\mathbb{F}$ is infinite or any sufficiently large field. Then $r(Y_1, \ldots, Y_n)$ is an invertible matrix in $M_k(\mathbb{F}(Y))$ where $Y_1, \ldots, Y_n$ are generic indeterminate matrices of dimension $k$.

Remark 12. Usually Theorem 11 is stated over infinite fields. However it can be adapted over any sufficiently large finite field $\mathbb{F}$ using the techniques in [25]. We briefly discuss it here. For details we refer the reader to ALGORITHM 1 in [25]. Define the field $\mathbb{F}'$ by adjoining a $k$th root $\zeta$ to $\mathbb{F}$ i.e. $\mathbb{F}' = \mathbb{F}[\zeta]$. Let $Z_1$ and $Z$ be distinct formal variables and consider the function field $\mathbb{F}'(Z_1, Z)$. Construct a $\mathbb{F}'(Z_1, Z)$-linear basis $\Gamma = \{C_1, \ldots, C_{k^2}\}$ of $M_k(\mathbb{F}'(Z_1, Z))$. It can be shown that $r$ is invertible on a generic linear combination of $\Gamma$. Now by a standard argument the generic variables can be fixed from $\mathbb{F}$ (assuming that $\mathbb{F}$ is sufficiently large) to obtain a matrix tuple $T$ such that $r(T)$ is invertible. This also implies that $r(Y)$ is invertible where $Y$ is a generic matrix tuple of dimension $k$.

2.2 Algebraic branching programs (ABPs)

Definition 13. An algebraic branching program (ABP) is a layered directed acyclic graph with one in-degree-0 vertex called source, and one out-degree-0 vertex called sink. Its vertex set is partitioned into layers 0, 1, \ldots, $d$, with directed edges only between adjacent layers $i$ to $i+1$. The source and the sink are in layers zero and $d$, respectively. Each edge is labeled by a linear form over $\mathbb{F}$ in variables $\{x_1, \ldots, x_n\}$. The polynomial computed by the ABP is the sum over all source-to-sink directed paths of the product of linear forms that label the edges of the path. The maximum number of nodes in any layer is called the width of the algebraic branching program. The size of the branching program is taken to be the total number of nodes.

Equivalently, an ABP of width $w$ and $d$ many layers can be defined as an entry of a product of $d$ many linear matrices of size at most $w$. Therefore, the polynomial $f$ computed by an ABP is of form $(M_1 \cdots M_d)_{i,j}$ for some $i, j \in [w]$. 
Proposition 14. An ABP of size $s$ has a linear pencil of size at most $2s$ from the following construction:

$$L_f = \begin{bmatrix} I_w & -M_1 & \cdots & -M_d \\ I_w & -M_2 & \cdots & \cdots \\ \vdots & \ddots & \ddots & \vdots \\ I_w & \cdots & -M_d & I_w \end{bmatrix}.$$ 

The ABP is computed in the upper right corner.

This construction is well-known and also used in [23].

2.3 Matrix Inverse

Let $P$ be a $2 \times 2$ block matrix shown below.

$$P = \begin{bmatrix} p_1 & p_2 \\ p_3 & p_4 \end{bmatrix}$$

where $p_1$ is invertible and $p_2$ and $p_3$ can be any rectangular matrices and $(p_4 - p_3p_1^{-1}p_2)$ is also invertible. Then we note that the inverse of $P$ has the following structure [23],

$$P^{-1} = \begin{bmatrix} p_1^{-1}(I + p_3(p_4 - p_3p_1^{-1}p_2)^{-1}p_3p_1^{-1}) & -p_1^{-1}p_2(p_4 - p_3p_1^{-1}p_2)^{-1}p_3p_1^{-1} \\ -(p_4 - p_3p_1^{-1}p_2)^{-1}p_3p_1^{-1} & (p_4 - p_3p_1^{-1}p_2)^{-1} \end{bmatrix} \tag{1}$$

If $p_3 = 0$, then $P^{-1}$ has a simpler structure.

$$P^{-1} = \begin{bmatrix} p_1^{-1} & -p_1^{-1}p_2p_4^{-1} \\ 0 & p_4^{-1} \end{bmatrix}. \tag{2}$$

Hrubeš and Wigderson use Equation 1 to compute each entry of the matrix inverse recursively by a small rational circuit.

Theorem 15 ([23, Theorem 2.4]). Each entry of the inverse of an $s \times s$ symbolic matrix is computable by a rational circuit of size $O(s^w)$ where $w$ is the exponent of matrix multiplication.

Remark 16. We observe that the same construction also yields a polynomial-size rational r-skewed circuit as defined in Definition 6 for the matrix inverse. Inspecting Equation 1, we just need to compute the entries of $p_1^{-1}$ and $(p_4 - p_3p_1^{-1}p_2)^{-1}$ and after that the remaining computation is straightforward. Notice that, in the composition step while replacing each $y_i$ by $g_i^{-1}$, Definition 6 allows any $g_i$ to be a sub-circuit of some $g_j$. Therefore, we can reuse the r-skewed circuit computing each entry of $p_1^{-1}$ and follow the same recursive construction to obtain a rational $r$-skewed circuit of size $O(s^w)$.

2.4 Recognizable series

A comprehensive treatment is in the book by Berstel and Reutenauer [8]. We will require the following concepts. Recall that $\mathbb{F}[[x]]$ is the formal power series ring over a field $\mathbb{F}$. A series $S$ in $\mathbb{F}[[x]]$ is recognizable if it has the following linear representation: for some integer $s$, there exists two column vectors $c, b \in \mathbb{F}^s$ and an $s \times s$ matrix $M$ whose entries are homogeneous linear forms over $x_1, \ldots, x_n$ i.e. $\sum_{i=1}^n a_i x_i$ such that $S = c^T \left( \sum_{k \geq 0} M^k \right) b$. Equivalently, $S = c^T (I - M)^{-1} b$. We say, $S$ has a representation $(c, M, b)$ of size $s^3$.

\footnote{In the language of weighted automata, the matrix $M$ is the transition matrix for the series $S$.}
The following theorem is a basic result in algebraic automata theory, generally attributed to Schützenberger. It has a simple linear algebraic proof [16, Corollary 8.3, Page 145].

**Theorem 17.** A recognizable series with representation \((c, M, b)\) of size \(s\) is nonzero if and only if \(c^t (\sum_{k \leq s-1} M^k) b\) is nonzero.

In this paper, the theorem is used to argue that the truncated series is computable by a small noncommutative ABP, thereby reducing zero-testing of recognizable series to the identity testing of noncommutative ABPs.

### 2.5 Matrix coefficient realization theory

We do not know any canonical form for noncommutative rational functions. However, if a noncommutative rational function is analytic (or defined) at a matrix point, then (matrix coefficient)-realization theory Volčič [34] offers a power series representation of the noncommutative rational function around that point. This is found useful in automata theory and control theory.

A generalized word or a generalized monomial in \(x_1, \ldots, x_n\) over the matrix algebra \(M_m(F)\) allows matrices to interleave between variables. More formally, a generalized word over \(M_m(F)\) is of the form \(a_0x_{i_1}a_1 \cdots a_{d-1}x_{i_d}a_d\), where \(a_i \in M_m(F)\). A generalized polynomial over \(M_m(F)\) is a finite sum of generalized monomials in the ring \(M_m(F)(x)\). Similarly, a generalized series over \(M_m(F)\) is an infinite sum of generalized monomials in the ring \(M_m(F)(\{x\})\).

Let \(I = \{e_{i,j}, 1 \leq i, j \leq m\}\) be the set of matrix units which forms a linear basis for \(M_m(F)\). A generalized monomial \(m\) of degree \(d\) over \(M_m(F)\) can be expressed as a linear combination of generalized monomials of the form \(e_{i_0,j_0}x_{k_1}e_{i_1,j_1} \cdots e_{i_d,j_d}x_{k_d}e_{i_{d+1},j_{d+1}}\) by expressing each matrix \(a\) occurring in \(m\) as an \(F\)-linear combination in the \(E\)-basis. Hence, we can express any generalized series \(S\) over \(M_m(F)\) as a sum of generalized monomials over only \(E\) and \(x\), which we call its canonical representation. We say the series (resp. polynomial) \(S\) is identically zero if and only if it is zero under such expansion i.e. the coefficient of each generalized monomial in the canonical representation is zero.

The evaluation of a generalized series over \(M_m(F)\) is defined on any \(k'm \times k'm\) matrix algebra for some integer \(k' \geq 1\) [34]. To match the dimension of the coefficient matrices with the matrix substitution, we use an inclusion map \(\iota : M_m(F) \rightarrow M_{k'm}(F)\), for example, \(\iota\) can be defined as \(\iota(a) = a \otimes I_{k'}\) or \(\iota(a) = I_{k'} \otimes a\). We now define the evaluation of a generalized series (resp. polynomial) over \(M_m(F)\) in the following way. Any \(d\)-generalized word \(a_0x_{i_1}a_1 \cdots a_{d-1}x_{i_d}a_d\) over \(M_m(F)\) on a matrix substitution \((p_1, \ldots, p_n) \in M_m^{k'm}(F)\) evaluates to

\[\iota(a_0)p_{i_1}\iota(a_1) \cdots \iota(a_{d-1})p_{i_d}\iota(a_d)\]

under some inclusion map \(\iota : M_m(F) \rightarrow M_{k'm}(F)\). In ring theory, all such inclusions are known to be compatible by the Skolem-Noether theorem [31, Theorem 3.1.2]. Therefore, if a series \(S\) is zero with respect to some inclusion map \(\iota : M_m(F) \rightarrow M_{k'm}(F)\), then it must be zero for any such inclusions. The equivalence of the two notions of zeroseness follows from the proof of [34, Proposition 3.13].

We now recall the definition of a recognizable generalized series from the same paper.

**Definition 18.** A generalized series \(S\) in \(M_m(F)(\{x\})\) is recognizable if it has the following linear representation. For some integer \(s\), there exists a row-tuple of matrices \(c \in (M_m(F))^{1 \times s}\), and \(b \in (M_m(F))^{s \times 1}\) and an \(s \times s\) matrix \(M\) whose entries are homogeneous generalized linear forms over \(x_1, \ldots, x_n\) i.e. \(\sum_{i=1}^n p_i x_i q_i\) where each \(p_i, q_i \in M_m(F)\) such that \(S = c(I - M)^{-1}b\). We say, \(S\) has a linear representation \((c, M, b)\) of size \(s\) over \(M_m(F)\).
On Identity Testing and Noncommutative Rank Computation

In [34], Volčič shows the following result.

**Theorem 19** ([34, Corollary 5.1, Proposition 3.13]). Given a noncommutative rational formula \( r \) of size \( s \) over \( x_1, \ldots, x_n \) and a matrix tuple \( p \in M_{m}^{n}(F) \) in the domain of definition of \( r \), \( r(x+p) \) is a recognizable generalized series with a representation of size at most \( 2s \) over \( M_{m}^{n}(F) \). Additionally, \( r(x) \) is zero in the free skew field if and only if \( r(x+p) \) is zero as a generalized series.

**Proof.** For the first part, see Corollary 5.1 and Remark 5.2 of [34].

To see the second part, let \( r(x) \) be zero in the free skew field. Then the fact that \( r(x+p) \) is a zero series follows from Proposition 3.13 of [34]. If \( r(x) \) is nonzero in the free skew field, then there exists a matrix tuple \((q_1, \ldots, q_n) \in M_{m}^{n}(F) \) such that \( r(q) \) is nonzero. W.l.o.g. we can assume \( l = k'm \) for some integer \( k' \). Fix an inclusion map \( \iota: M_{m}(F) \rightarrow M_{k'm}(F) \). Define a matrix tuple \((q'_1, \ldots, q'_n) \in M_{k'm}^{n}(F) \) such that \( q'_i = q_i - \iota(p_i) \). Therefore, the series \( r(x+p) \) on \((q'_1, \ldots, q'_n) \) evaluates to \( r(g) \) under the inclusion map \( \iota \), hence nonzero [34, Remark 5.2]. Therefore, \( r(x+p) \) is also nonzero.

**Remark 20.** More explicitly we can say the following which is already outlined in [34, Section 5]. For any inclusion map \( \iota: M_{m}(F) \rightarrow M_{k'm}(F) \)

\[
r(g + \iota(p)) = \iota(e) \left( I_{2sk'm} - \sum_{j=1}^{n} \iota(A^{x_j})(g) \right)^{-1} \iota(b).
\]

We also note down a few basic facts. The following is easy to show and also noted in [34].

**Fact 21.** Let \( r(x+p) \) be a generalized series where \( p \) consists of matrices in \( M_{m}(F) \). If we replace each \( x_i \) by a generic matrix over noncommuting variables \((y_{i,j,k})_{1 \leq i,j,k \leq m} \), then we get a nonzero matrix over the \( y \) variables. More precisely, the map \( \psi(x_i) = (y_{i,j,k})_{1 \leq i,j,k \leq m} \) is identity preserving.

Another easy fact is the following.

**Fact 22.** Let \( r(x+p) \) has a linear representation \( c(I - M)^{-1}b \) of size \( s \). Then each entry of \( r(\psi(x)+p) \) is a recognizable series with transition matrix \( M(\psi(x_1), \ldots, \psi(x_n)) \) of size \( sm \).

More precisely, the \((i,j)^{th} \) entry of \( r(\psi(x)+p) \) has a representation \((e_i, M(\psi(x)), b_j) \) where \( e_i \) and \( b_j \) are the \( i^{th} \) row and \( j^{th} \) column of \( c \) and \( b \) respectively.

## 3 Derandomization of RIT from the Hardness of Polynomial Identities

In this section, we present a new approach to derandomize (almost general) RIT efficiently in the black-box setting and prove Theorem 4. Given a noncommutative polynomial \( P(x_1, \ldots, x_n) \in F(x_1, \ldots, x_n) \), there is a well-known trick to reduce the identity testing of \( P \) to the identity testing of a bivariate polynomial \( P'(y_0, y_1) \) over the noncommuting variables \( y_0, y_1 \) by the substitution \( x_i \leftarrow y_0 y_i^t y_0 \) for \( 1 \leq i \leq n \).

For a rational formula \( r(x) \), such a variable reduction step preserving identity is not immediate. Our first result in this section reduces the identity testing of an \( n \)-variate rational formula of inversion height \( h \) to the identity testing of a rational formula of inversion height \( h \) over \( 2(h+1) \) variables. But before that we record a simple fact.

**Fact 23.** Given any rational formula \( r' \) of of inversion height at most \( h - 1 \) and size at most \( s \), if we can find a matrix tuple such that \( r' \) is invertible on that matrix tuple, then for a rational formula \( r \) of size at most \( s \) and inversion height \( h \), we can find a matrix tuple where \( r \) is defined.
Proof. Let $F$ be the collection of all those inverse gates in the formula $\tau$ such that for every $g \in F$, the path from the root to $g$ does not contain any inverse gate. For each $g_i \in F$, let $h_i$ be the sub-formula input to $g_i$. Consider the formula $\tau' = h_1 h_2 \cdots h_k$ (where $k = |F|$) which is of size at most $s$ since for each $i$ and $j$, $h_i$ and $h_j$ are disjoint. Clearly, $\tau'$ is of inversion height at most $h - 1$. So if we find a point $q$ such that $\tau'(q)$ is invertible then $\tau$ is defined at that point $q$. □

Now we state and prove the variable reduction lemma for rational formulas.

Lemma 24. Let $\tau(x_1, \ldots, x_n)$ be a rational formula of inversion height $h$. Then, there exists a $2(h + 1)$ variate rational formula $\tau'$ of inversion height $h$ over the variables $\{y_{j0}, y_{j1} : 0 \leq j \leq h\}$ such that $\tau$ is zero in $F(x)$ if and only if $\tau'$ is zero in $F(y)$. Moreover, $\tau'$ is obtained from $\tau$ by substituting $x_i$ by $\sum_{j=0}^{h} y_{j0} y_{j1}^{i} y_{j0}$ for $1 \leq i \leq n$.

Proof. The proof is by induction on the inversion height $h$. In fact we use a stronger inductive hypothesis: For every nonzero rational formula $\tau$ of inversion height $h$, there also exists a matrix tuple $(p_1, \ldots, p_n)$ and a collection of matrices $\{q_{00}, q_{01}, \ldots, q_{h0}, q_{h1}\}$ such that $\tau(p_1, \ldots, p_n)$ is invertible and for each $i \in [n]$, $p_i = \sum_{j=0}^{h} q_{j0} q_{j1}^{i} y_{j0}$.

It is true for noncommutative polynomials for which $h = 0$. It is already mentioned that the substitution $x_i \leftarrow y_{00} y_{01}^{i}$ reduces the identity testing of $P(x)$ to the identity testing of $P'(y_0, y_1)$. Moreover, by Theorem 11, we know that we can find matrices $q_0, q_1$ such that the bivariate polynomial $P'(q_0, q_1)$ evaluates to an invertible matrix. Since $P'(q_0 y_0, q_0 y_1$, $q_0 y_0^{i}, y_{00}, q_{h0}) = P'(q_0, q_1)$, we establish the base case of the induction.

Inductively, suppose that it is true for any formula of inversion height $h - 1$. Now consider a nonzero rational formula $\tau(x_1, \ldots, x_n)$ of inversion height $h$. From the inductive hypothesis and Fact 23, there exists a matrix tuple $(p_1, \ldots, p_n)$ and a collection of matrices $\{q_{00}, \ldots, q_{(h-1)0}, q_{01}, \ldots, q_{(h-1)1}\}$ such that $\tau(p_1, \ldots, p_n)$ is defined and for each $i \in [n]$, $p_i = \sum_{j=0}^{h-1} q_{j0} q_{j1}^{i} y_{j0}$. Let the dimension of each $p_i$ be $m$. Therefore, $\tau(x + p)$ is also a nonzero generalized series by Theorem 19. Replacing each $x_i$ by $y_0 y_1^{i} y_{00}$, we obtain a nonzero bivariate generalized series and suppose it is nonzero for $y_0 = q_{00}$ and $y_1 = q_{01}$ of some dimension $km$ for an integer $k$. Notice from Section 2 that a generalized series is zero if and only if the coefficient of every monomial in the canonical representation is zero. Therefore the bivariate substitution $x_i \to y_0 y_1^{i} y_{00}$ preserves the nonzeroness of a generalized series. Therefore,

$$\tau(q_{00} q_{01} q_{00}^{i} + \epsilon(p_1), \ldots, q_{00} q_{01}^{n} q_{00} + \epsilon(p_n))$$

is also nonzero. Notice that, $\epsilon(p_i) = \sum_{j=0}^{h-1} \epsilon(q_{j0}) (\epsilon(q_{j1}))^{i} (\epsilon(q_{j0}))$ for the inclusion map $\epsilon$ from $\mathbb{M}_m(F) \to \mathbb{M}_{km}(F)$. We can now define $\tau'$ substituting each $x_i$ in $\tau$ by $\sum_{j=0}^{h} y_{j0} y_{j1}^{i} y_{j0}$. Clearly, $\tau'$ is nonzero. By Theorem 11, $\tau'$ is also invertible for some matrix tuple $q$ of same dimension. Hence $\tau(p_1, \ldots, p_n)$ is invertible for $p_i = \sum_{j=0}^{h} q_{j0} q_{j1}^{i} q_{j0}$. □

Next we show that if Conjecture 2 is true then any rational formula of size $s$ and inversion height $h \leq \beta(\log s/\log \log s)$ for $\beta \in (0, 1)$, is nonzero on a matrix tuple of dimension $(\gamma \log s)^{h+1}$ for some constant $\gamma$.

Lemma 25. Let $\tau(x_1, \ldots, x_n)$ be a nonzero rational formula of size $s$ and inversion height $h \leq \beta(\log s/\log \log s)$ for any constant $0 < \beta < 1$. Then, Conjecture 2 implies that there is a matrix tuple $(p_1, \ldots, p_n) \in \mathbb{M}_m^n(F)$ such that $\tau(p_1, \ldots, p_n)$ is invertible and $m = (\gamma \log s)^{h+1}$ for some constant $\gamma > 1$. □
Proof. The proof is by induction on \( h \). For the base case \( h = 0 \), Conjecture 2 implies that the noncommutative formula is nonzero on generic \( c \log s \) (for some constant \( c \)) dimensional matrix tuple \((Z_1, \ldots, Z_n)\) where \( Z_i = (z_{i,j,k}^{(i)})_{1 \leq j,k \leq c \log s} \). Also Theorem 11 says that the formula evaluates to an invertible matrix \( M(Z) \) on substituting \( x_i \) by \( Z_i \). Now using standard idea, random substitution to the variables in \( Z_1, \ldots, Z_n \) yields such a matrix tuple.

Inductively assume that we have already proved the dimension bound on the witness of the invertible image for rational formulas of inversion height at most \( h - 1 \). Let the dimension of the matrices be \( d_{h-1} \). Now given a rational formula \( \tau \) of size \( s \) and inversion height \( h \), observe that \( \tau \) is defined on some \( d_{h-1} \times d_{h-1} \) matrix tuple \( q \) using Fact 23.

Then by Theorem 19, \( \tau(x + q) \) can be represented by a recognizable generalized series of size at most \( 2s \) such that \( \tau(x) \) is nonzero if and only if \( \tau(x + q) \) is nonzero. Using Fact 21, apply the \( \psi \) map on the variables such that \( \psi(x_i) \) substitutes the variable \( x_i \) by a matrix of fresh noncommuting variables \( z_{j,k}^{(i)} \) for \( 1 \leq j, k \leq d_{h-1} \).

Using Fact 22, observe that we get a matrix of recognizable series and each such recognizable series can be represented by an automaton of size at most \( s \leq 2sd_{h-1} \). So w.l.o.g., let the series be \( S_{1,1} \) computed at \((1,1)^{th} \) entry is nonzero. Let the transition matrix for \( S_{1,1} \) is \( M_{1,1} \). Then using Theorem 17, the truncated finite series \( \tilde{S}_{1,1} = \epsilon^t \left( \sum_{k \leq s-1} M_{1,1}^k \right) b \) is nonzero, which is a noncommutative ABP.

If Conjecture 2 is true then \( \tilde{S}_{1,1} \) will be nonvanishing on a matrix tuple \( p \) of dimension \( O(\log s) \). Now by the following simple scaling trick, we show that the infinite series \( S_{1,1} \) is nonzero at a matrix tuple of dimension \( c \log s \).

\[ \triangleright \text{Claim 26. We can find a matrix tuple } p' \text{ which is a scalar multiple of } p \text{ such that } S_{1,1}(p') \text{ is nonzero.} \]

\[ \textbf{Proof.} \text{ Let } \tau \text{ be a commutative variable and consider the matrix tuple, } \]

\[ \tau p = (\tau p_{1,1}^{(1)}, \ldots, \tau p_{d_{h-1}, d_{h-1}}^{(1)}, \ldots, \tau p_{1,1}^{(n)}, \ldots, \tau p_{d_{h-1}, d_{h-1}}^{(n)}). \]

Observe that \( M_{1,1}(\tau p) = \tau M_{1,1}(p) \). From the definition of the series \( S_{1,1} \),

\[ S_{1,1}(z) = \tilde{S}_{1,1}(z) + \sum_{t \geq s} \epsilon^t M_{1,1}^t b. \]

Let \( d \) be the dimension of the matrices in the tuple \( p \). We now evaluate \( S_{1,1} \) at \( \tau p \) to get the following:

\[ S_{1,1}(\tau p) = \tilde{S}_{1,1}(\tau p) + \sum_{t \geq s} \epsilon^t \cdot ((\epsilon \otimes I_d)^t \cdot M_{1,1}^t(p) \cdot (b \otimes I_d)). \]

Since \( \tilde{S}_{1,1}(p) \neq 0 \), we have that \( S_{1,1}(\tau p) \) evaluates to a nonzero matrix whose entries are power series in the variable \( \tau \).

It is also true that \( S_{1,1}(\tau p) = (\epsilon \otimes I_d)^t \cdot (I - M_{1,1}(\tau p))^{-1} \cdot (b \otimes I_d) \) which is rational expression in \( \tau \) where the degrees of the numerator and denominator polynomials are bounded by \( \text{poly}(s, d) \). Hence we need to avoid only \( \text{poly}(s, d) \) values for \( \tau \) such that \( S_{1,1}(\tau p) \) is defined and nonzero.

The above argument shows that for a specific value \( \tau_0 \) for the parameter \( \tau \), the generalized series \( \tau(x + q) \) evaluates to nonzero on a matrix tuple \((N_1(\tau_0) + \iota(q_1), \ldots, N_n(\tau_0) + \iota(q_n))\) where \( N_i \) is obtained from the matrix \((z_{j,k}^{(i)})_{1 \leq j,k \leq d_{h-1}}\) by substituting the variables \((z_{j,k}^{(i)})_{1 \leq j,k \leq d_{h-1}}\) by \( \tau_0 p_{j,k}^{(i)} \). Also \( \iota \) is the inclusion map \( \iota : M_{d_{h-1}}(\mathbb{F}) \rightarrow M_{d_{h-1}}(\mathbb{F}) \) defined as \( \iota(q_i) = q_i \otimes I_d. \)
Hence $r$ is nonzero on generic matrix tuples of dimension $d_h = \sum d - 1 \leq \sum d - 1 \log (s d - 1)$. Inductively assume that $d_h - 1 \leq (2 \log s)^h$. Since $h \leq \beta (\log s / \log \log s)$, we can observe that $s \geq d_h - 1$. Using this we get that $d_h \leq c (2 \log s)^h \log (s^2)$ and that yields $d_h \leq (2 \log s)^{h+1}$.

We take $\gamma = 2c$.

Therefore by Theorem 11 $r(x)$ evaluates to an invertible matrix on substituting $x_i$ by generic matrices of dimension $(\gamma \log s)^{h+1}$.

Now we are ready to show that if Conjecture 2 is true, then we can find a subexponential-size hitting set for rational formulas of size $s$ and inversion height up to $c' (\log s / \log \log s)$ for a suitable constant $c'$ that depends on the exponent of the subexponential function.

Proof of Theorem 4. Let $r(x_1, \ldots, x_n)$ be a rational formula of inversion height $h$ and size $s$. Consider, $r'(y_0, y_1, \ldots, y_{h0}, y_{h1})$ obtained from $r$ by substituting $x_i$ by $\sum_{j=0}^h y_{j0} y_{j1}^i y_{j0}$ for $1 \leq i \leq n$. From Lemma 24, we know that $r(x)$ is nonzero if and only if $r'(y)$ is nonzero. Moreover, $r'$ has a rational formula of size at most $s'$ which is of $O(sn \log \log s)$. Therefore, $r'$ must be invertible on $d_h \times d_h$ generic matrix substitution where $d_h \leq (\gamma \log s')^{h+1}$ from Lemma 25. Using Proposition 10, we know that $r'$ has a linear pencil $L'$ of size at most $2s'$. W.l.o.g., assume that $r'$ is computed at the $(1,1)$th entry of $L'$.

Hence, if we substitute the variables $y_{00}, y_{01}, \ldots, y_{h0}, y_{h1}$ by $d_h \times d_h$ generic matrices $\{Z^{(i,0)}, Z^{(i,1)} : 0 \leq i \leq h\}$ (over commuting variables), the $(1,1)$th block of $L'$ will be of form $M'(Z) / \det(L'(Z))$ where $\det(L'(Z))$ is a polynomial of degree at most $2s' (\gamma \log s')^{h+1}$. Further, each entry of the matrix $M'$ is a cofactor of $L'(Z)$ and therefore it is a polynomial over the $Z$ variables of degree at most $2s' (\gamma \log s')^{h+1}$. This shows that $\det(M'(Z))$ is a nonzero polynomial of degree at most $2s' (\gamma \log s')^{2h+2}$. The sparsity of $\det(L'(Z))$ and $\det(M'(Z))$ are bounded by

$$\kappa = (s' (\gamma \log s)^{2h+2}) O(h (\gamma \log s)^{2h+2}).$$

Now we can use standard sparse polynomial hitting set for $\kappa$-sparse polynomials to hit both the polynomials [28]. This gives us a strong hitting set $H'$ for $r'$. Consequently, we get a strong hitting set of same size for $r$ by using the substitutions of $x_i$ variables by the $y_{00}, y_{01}, \ldots, y_{h0}, y_{h1}$ described in Lemma 24. More formally, we define

$$H_{n,h,s} = \{ (q_1, \ldots, q_n) : q \in H' ; p_i = \sum_{j=0}^h q_{j0} q_{j1}^i q_{j0} \}.$$

An immediate corollary is the following.

Corollary 27. The hitting set size and the construction time is $s(\log s)^{O(1)}$ for $h = O(1)$. If we want to maintain a subexponential-size hitting set of size $2s^\delta$ for $\delta \in (0,1)$, then $h$ can be taken to be at most $c_3 \left( \frac{\log s}{\log \log s} \right)$ where $c_3$ is a constant depending on $\delta$.

4 Computing the Matrix Rank over the Free Skew Field

In this section, we give an efficient algorithm to compute the rank of any matrix over the free skew field whose entries are rational functions with small linear pencils. Additionally, we output a (witness) matrix tuple on which the rank is achieved. This is done in two steps. We first introduce a blow-up definition for matrix rank over the free skew field extending the results for linear pencils and show that it is equivalent to the usual definition of noncommutative rank. Next, we show an efficient reduction from rank computation over the free skew field to the linear case in Section 4.1. A discussion on the blow-up definition of noncommutative rank can be found in the full version [2].
4.1 The Rank Computation

In this section, we prove Theorem 5. The idea is to reduce rank computation of a matrix with skew field entries from \( F \langle \bar{x} \rangle \) to rank computation of a linear matrix over \( \bar{x} \) incurring a small blow-up in the matrix size.

**Lemma 28.** Let \( P \in F \langle \bar{x} \rangle^{m \times m} \) such that,

\[
P = \begin{bmatrix} A & B \\ C & D \end{bmatrix},
\]

where \( A \in F \langle \bar{x} \rangle^{r \times r} \) is invertible. Then,

\[
\text{ncrank}(P) = r + \text{ncrank}(D - CA^{-1}B).
\]

**Proof.** If \( Q \) is an \( n \times n \) invertible matrix over \( F \langle \bar{x} \rangle \) then \n\[
\text{ncrank}(QP) = \text{ncrank}(PQ) = \text{ncrank}(P).
\]

For if \( P = MN \) then \( QP = (QM)N \) and if \( PQ = MN \) then \( P = (Q^{-1}M)N \). Similarly for \( PQ \).

The matrix

\[
\begin{bmatrix}
A^{-1} & 0 \\
0 & I_{m-r}
\end{bmatrix}
\]

is full rank. Similarly, the matrix

\[
\begin{bmatrix}
I_r & 0 \\
-C & I_{m-r}
\end{bmatrix}
\]

is full rank because

\[
\begin{bmatrix}
I_r & 0 \\
-C & I_{m-r}
\end{bmatrix} \begin{bmatrix}
I_r & 0 \\
C & I_{m-r}
\end{bmatrix} = \begin{bmatrix}
I_r & 0 \\
0 & I_{m-r}
\end{bmatrix}.
\]

Hence, \( \text{ncrank}(P) \) equals \( \text{ncrank}(R) \) where

\[
R = \begin{bmatrix}
I_r & 0 \\
-C & I_{m-r}
\end{bmatrix} \begin{bmatrix} A^{-1} & 0 \\ 0 & I_{m-r} \end{bmatrix} \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I_r & A^{-1}B \\ 0 & D - CA^{-1}B \end{bmatrix}
\]

Post-multiplying by the invertible matrix

\[
\begin{bmatrix}
I_r & -A^{-1}B \\
0 & I_{m-r}
\end{bmatrix}
\]

we obtain

\[
\begin{bmatrix}
I_r & 0 \\ 0 & D - CA^{-1}B
\end{bmatrix}.
\]

It is easy to see that its inner rank is \( r + \text{ncrank}(D - CA^{-1}B) \). ▶

Next, we relate the noncommutative rank of a matrix with skew field entries with small linear pencils to the noncommutative rank of a linear matrix.

**Lemma 29.** Let \( M \in F \langle \bar{x} \rangle^{m \times m} \) be a matrix whose \((i,j)^{th}\) entry \( g_{ij} \) is computed as the \((1,1)^{th}\) entry of the inverse of a linear pencil \( L_{ij} \) of size at most \( s \), for each \( i \) and \( j \). Then, one can construct a linear pencil \( L \) of size \( m^2s + m \) such that,

\[
\text{ncrank}(L) = m^2s + \text{ncrank}(M).
\]
Proof. We first describe the construction of the linear pencil $L$ and then argue the correctness. W.l.o.g. we may assume that each linear matrix $L_{ij}$ is $s \times s$ (by padding it, if required, with an identity matrix of suitable size).

Let

$$L = \begin{bmatrix} L_{11} & 0 & \cdots & 0 & B_{11} \\ 0 & L_{12} & \cdots & 0 & B_{12} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & L_{mm} & B_{mm} \\ -C_{11} & -C_{12} & \cdots & -C_{mm} & 0 \end{bmatrix},$$

(3)

where each $C_{ij}$ is an $m \times s$ and $B_{ij}$ is an $s \times m$ rectangular matrix defined below. Let $e_i$ denote the column vector with 1 in the $i^{th}$ entry and the remaining entries are zero. We define

$$C_{ij} = \begin{bmatrix} e_i \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \text{and} \quad B_{ij} = \begin{bmatrix} e_i^T \\ 0 \\ \vdots \\ 0 \end{bmatrix}.$$

To argue the correctness of the construction, we write $L$ as a $2 \times 2$ block matrix. As each $L_{ij}$ is invertible (otherwise $g_{ij}$ would not be defined), the top-left block entry is invertible. Therefore, we can find two invertible matrices $U, V$ implementing the required row and column operations such that,

$$L = U \begin{bmatrix} L_{11} & 0 & \cdots & 0 & 0 \\ 0 & L_{12} & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & L_{mm} & 0 \\ 0 & 0 & \cdots & 0 & D \end{bmatrix} V,$$

for some $m \times m$ matrix $\tilde{D}$.

▷ Claim 30. The matrix $\tilde{D}$ is exactly the input matrix $M$.

Proof. From the $2 \times 2$ block decomposition we can write,

$$\tilde{D} = [C_{11}C_{12}\cdots C_{mm}] \begin{bmatrix} L_{11}^{-1} & 0 & \cdots & 0 & \sum_{i,j} C_{ij} L_{ij}^{-1} B_{ij} \\ 0 & L_{12}^{-1} & \cdots & 0 & \vdots \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & L_{mm}^{-1} & \vdots \\ 0 & 0 & \cdots & 0 & B_{mm} \end{bmatrix}.$$

Observe that, for each $i, j$, $C_{ij} L_{ij}^{-1} B_{ij}$ is an $m \times m$ matrix with $g_{ij}$ as the $(i, j)$\textsuperscript{th} entry and remaining entries are 0. Hence, $\tilde{D} = M$. Notice that the top-left block of $L$ in Equation 3 is invertible as for each $i, j \in [m]$, $L_{ij}$ is invertible. Now the proof follows from Lemma 28.

Proof of Theorem 5. For any matrix $M = (g_{i,j})_{m \times m}$ such that for each $i, j \in [m]$, $g_{ij}$ in $\mathbb{F}<x_1, \ldots, x_n>$ has a linear pencil of size at most $s$, construct a linear matrix $L$ of size $m^2 s + m$ from the previous lemma. We can now compute the noncommutative rank of $L$
using the algorithm of [25] in deterministic poly\((s,m,n)\)-time. Let the rank be \( r \). We now output \( r - m^2 s \) to be the noncommutative rank of \( M \). The correctness of the algorithm follows from Lemma 29.

By the equivalence of the inner rank and blow-up rank, we know that \( \text{ncrank}^*(M) = r - m^2 s \). Now we use the algorithm in [25] to compute a matrix tuple \( p \in M_d(\mathbb{F}) \) such that the rank of \( L(p) = rd \) for some \( d = O(m^2 s) \). Clearly \( \text{rank}(M(p)) = (r - m^2 s)d \). Therefore, the matrix tuple \( p \) is also a witness of the rank of \( M \).

\[ \begin{align*}
\end{align*} \]

### 5 Efficient Linear Pencils for Inversely Disjoint r-Skewed Circuits

We now prove that an inversely disjoint r-skewed circuit of size \( s \) has a linear pencil representation of size \( O(s^2) \). We first prove a more general result, a composition lemma for linear pencils which implies Theorem 8.

**Lemma 31.** Let \( L \) be an \( s \times s \) linear pencil over \( x_1, \ldots, x_n \) and \( y_1, \ldots, y_m \). Let \( f_{i,j} = (L^{-1})_{i,j} \) for \( i, j \in [s] \). Let \( g_1, \ldots, g_m \) be rational functions over \( x_1, \ldots, x_n \) such that each \( g_k \) has a linear pencil \( L_k \) of size at most \( s_k \). Then we can construct a single linear pencil \( L \) of size \( \sum_{i=1}^m s_i + m + 2s^2 + s \) in \( \text{poly}(s_1, \ldots, s_m, s, m, n) \)-time such that

\[ (\tilde{L}^{-1})_{2s^2+1+2s^2+1} i,j = f_{i,j}(x, g_1^{-1}, \ldots, g_m^{-1}) \quad \text{for} \quad i, j \in [s], \quad \text{where} \quad \tilde{s} = \sum_{i=1}^m s_i + m. \]

**Proof.** For each \( i, j \in [s] \), \( f_{i,j} = (L^{-1}(x, y_1, \ldots, y_m))_{(i,j)} \). We now define for each \( i, j \in [s] \), \( h_{i,j} = f_{i,j}(x, g_1^{-1}, \ldots, g_m^{-1}) \). As the variables \( y_1, \ldots, y_m \) are indeterminates, we can rewrite each \( h_{i,j} \) as the following:

\[ h_{i,j} = (L^{-1}(x, g_1^{-1}, \ldots, g_m^{-1}))_{(i,j)}. \]

We first describe the construction of the linear pencil \( \tilde{L} \) and then prove the correctness of the construction. Let \( \tilde{L} \) be a linear pencil over \( x_1, \ldots, x_n \) of size \( \tilde{s} \) where for each \( k \in [m] \), there exists \( i_k, j_k \in [s] \) such that \( \tilde{g}_k^{-1} = (\tilde{L}^{-1})_{i_k,j_k} \). The description of \( \tilde{L} \) is given later.

Let us first define two \( s \times s \) linear pencils \( L' \) and \( L'' \) as follows. Fix \( i, j \in [s] \). Let \( (L)_{i,j} = \alpha_0 + \sum_{k=1}^m \alpha_{k,i,j}x_k + \sum_{k=1}^m \beta_{k,i,j}y_k \). Write \( L = L' + L'' \) such that \( (L')_{i,j} = \alpha_0 + \sum_{k=1}^m \alpha_{k,i,j}x_k \) and \( (L'')_{i,j} = \sum_{k=1}^m \beta_{k,i,j}y_k \). We now define \( \tilde{L} \) as a \( 4 \times 4 \) block linear matrix of size \( \tilde{s} + 2s^2 + s \),

\[ \tilde{L} = \begin{bmatrix}
I_{s^2} & A_1 & 0 & 0 \\
0 & L & A_2 & 0 \\
0 & 0 & I_{s^2} & A_3 \\
A_4 & 0 & 0 & L'
\end{bmatrix}, \quad \text{(4)} \]

where \( I_{s^2} \) is the identity matrix of size \( s^2 \) and \( A_1, A_2, A_3 \) and \( A_4 \) are some rectangular matrices of dimension \( s^2 \times \tilde{s}, \tilde{s} \times s^2, s^2 \times s \) and \( s \times s^2 \) respectively. We now define the construction \( A_1, A_2, A_3 \) and \( A_4 \). Subsequently in this proof \( I \) is used for \( I_{s^2} \).

Let \( \tilde{L}_1 = \begin{bmatrix} I & A_1 \\ 0 & L \end{bmatrix} \). Then \( \tilde{L}_1^{-1} = \begin{bmatrix} I & -A_1\tilde{L}^{-1} \\ 0 & L^{-1} \end{bmatrix} \).

We now consider the top-left \( 3 \times 3 \) block matrix.

Let \( \tilde{L}_2 = \begin{bmatrix} I & A_1 \\ 0 & \tilde{L} & A_2 \end{bmatrix} \). Then \( \tilde{L}_2^{-1} = \begin{bmatrix} \tilde{L}^{-1} & B_1 \\ 0 & I \end{bmatrix} \).
We now define, for each $S$. Simplifying further, here the vectors noncommutative ABP and the theorem holds by Proposition 14. by induction on the inversion height $h$ and of size $s$. Easy to see that for each $(i,j) \in [s] \times [s]$ and $k \in [m]$, $(A_1)_{(i,j),i_k} = \beta_{k,i,j}$, $(A_2)_{j_k,(i,j)} = 1$ and the other entries are zero. Then,

$$(B_2)_{(i,j),(i,j)} = \sum_{k=1}^m (A_1)_{(i,j),i_k} (\tilde{L}^{-1})_{i_k,j_k} (A_2)_{j_k,(i,j)} = \sum_{k=1}^m \beta_{k,i,j} g_k^{-1}.$$  

We now define, for each $i,j \in [s]$, $(A_4)_{i,(i,j)} = -1$ and 0 otherwise and $(A_3)_{(i,j),j} = 1$ and 0 otherwise. Since

$$\tilde{L} = \begin{bmatrix} I & A_1 & 0 & 0 \\ 0 & \tilde{L} & A_2 & 0 \\ 0 & 0 & I & A_2 \\ A_4 & 0 & 0 & \bar{L} \end{bmatrix},$$  

Now, $\tilde{L}^{-1} = \begin{bmatrix} * & * \\ * & B_3 \end{bmatrix}$, where, $B_3 = \begin{bmatrix} L' - (A_4 & 0 & 0) \tilde{L}_2^{-1} (0 & 0 \tilde{A_3}) \end{bmatrix}^{-1}$. Simplifying further,

$$B_3 = (L' - A_4B_2A_3)^{-1} = L^{-1}(x, g_1^{-1}, \ldots, g_m^{-1}).$$  

Therefore, for each $i,j \in [s]$, $(B_3)_{i,j} = (L^{-1}(x, g_1^{-1}, \ldots, g_m^{-1}))_{i,j} = h_{i,j}$.  

Now we construct the linear pencil $\hat{L}$ of size $s = \sum_{k=1}^m s_k + m$. For $k \in [m]$, let there are indices $i_k', j_k' \in [s_k]$ such that $g_k = (L_k^{-1})_{i_k', j_k'}$. We now define for each $k \in [m]$,  

$$\hat{L}_k := \begin{bmatrix} L_k & e_{i_k'} & 0 \\ -e_{j_k'} & 0 \end{bmatrix}.$$  

Here the vectors $e_i$ are the unit vector. The construction of $\hat{L}$ is now as follows:

$$\hat{L} = \begin{bmatrix} \hat{L}_1 & 0 & \ldots & 0 \\ 0 & \hat{L}_2 & \ldots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \ldots & \hat{L}_m \end{bmatrix}.$$  

Considering the $\hat{L}$ as an $m \times m$ block matrix where the $i^{th}$ block is of size $s_i + 1$, it is easy to see that for each $k \in [m]$, the bottom-right corner entry of the $k^{th}$ block of $\hat{L}^{-1}$ is $g_k^{-1}$. To see this apply Equation 1 with $p_4 = 0$.  

Now the proof of Theorem 8 follows easily from Lemma 31. **Proof of Theorem 8.** We show that inversely disjoint $r$-skewed rational functions of height $h$ and of size $s$ have linear pencils of size at most $cs^2$ for some constant $c$. We prove it by induction on the inversion height $h$. For the base case $h = 0$, the input circuit is a noncommutative ABP and the theorem holds by Proposition 14.
Let \( f(\bar{x}, g_1^{-1}, \ldots, g_m^{-1}) \) be an input inversely disjoint \( r \)-skewed rational function of height \( h \) computed by the circuit \( C' \). Replacing \( g_i^{-1} \) by new variable \( y_i \) we get a noncommutative ABP \( C'(\bar{x}, y) \) of size \( s' \leq s \). Again by Proposition 14, \( C' \) can be represented by a linear pencil of size at most \( 2s' \). Let \( g_1, \ldots, g_m \) are computed by inversely disjoint \( r \)-skewed circuits of size \( s_1, \ldots, s_m \) and inversion heights \( \leq h - 1 \). By the inductive hypothesis each \( g_k \) is computable by a linear pencil of size at most \( cs_k^2 \).

Hence by Lemma 31, there is a linear pencil of size \( S \) representing \( C'(\bar{x}, g_1^{-1}, \ldots, g_m^{-1}) \) which satisfies the following condition.

\[
S \leq c \sum_{k=1}^{m} s_k^2 + m + 8s'^2 + 2s'.
\]

Simplifying further,

\[
S \leq c \left( \sum_{k=1}^{m} s_k^2 + m + s'^2 \right),
\]

for sufficiently large \( c \). Since the sub-circuits for \( g_1, \ldots, g_m \) are disjoint, we get that \((\sum_{k=1}^{m} s_k^2 + m + s'^2) \leq (\sum_{k=1}^{m} s_k + m + s')^2 \leq s^2 \). So, \( S \leq cs^2 \) for some large constant \( c \).

We now prove the following property of the linear pencil constructed in Theorem 8.

**Proposition 32.** For any inversely disjoint rational \( r \)-skewed circuit computing \( \tau \in \mathbb{F} \langle \bar{x} \rangle \) and a tuple of matrix \( p \in M_{2n}(\mathbb{F}) \) for some finite \( m \), the following are equivalent.

1. \( \tau \) is defined at \( p \).
2. For every gate \( u \) which is an output gate or a child of an inverse gate, the pencil constructed in Theorem 8 corresponding to the rational expression computed at \( u \) is invertible at \( p \).

The proof of Proposition 32 can be found in the full version [2].

**Proof of Corollary 9.** Let \( \tau(\bar{x}, g_1^{-1}, \ldots, g_m^{-1}) \) be the input inversely disjoint \( r \)-skewed circuit of size \( s \). By Theorem 8, we construct a linear pencil \( \tilde{L} \) of size \( O(s^2) \) for \( \tau^{-1} \). Now by Proposition 32, \( \tau^{-1} \) is defined at \( p \) if and only if \( L(p) \) is invertible. But \( \tau \) is nonzero if and only if \( \tau^{-1} \) is defined [1]. So for nonzero testing of \( \tau \), it is enough to apply the singularity testing algorithms in [25] on the linear pencil \( \tilde{L} \) in white-box case. For the black-box case one can use the algorithm in [14]. In fact the result in [25] also gives the dimension upper bound of \( O(s^2) \) for the tensoring matrices on which \( \tilde{L} \) should be tested for singularity. This also leads to randomized polynomial-time black-box algorithm that simply substitutes the variables randomly from matrices of dimension \( O(s^2) \) over sufficiently large fields.

**Future Directions**

Our work raises the following questions for further research:

- The most important question is to obtain an *unconditional derandomization* of the black-box RIT problem. The current best known result is a quasipolynomial-time black-box RIT algorithm for rational formulas of inversion height at most two [3].
- Theorem 4 opens up a new motivation to further study the Conjecture 2. In [5], it is shown that a nonzero noncommutative polynomial of sparsity \( s \) can not be an identity for some \( k = O(\log s) \) dimensional matrix algebra. This solves a special case of the conjecture and the proof uses automata theoretic ideas very crucially. Can we improve these techniques to settle the conjecture completely?
The effective use of Higman’s trick has found new applications in randomized polynomial-time factorization algorithm for noncommutative formulas [4]. The proof of Theorem 5 does not use Higman’s trick. It would be interesting to see whether such ideas can be applied elsewhere.

Can we exactly characterize (up to a polynomial-size equivalence) the expressive power of linear pencil representations for some sub-class of rational circuits? In this paper, we show that inversely disjoint r-skewed circuits have polynomial-size linear pencils. This gives ID-R-rSC ⊆ LR. It would be very interesting to prove that rational r-skewed circuits can be expressed by polynomial-size linear pencils. In other words, prove that R-rSC = LR.

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