Computing Solutions of Linear Mahler Equations

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Abstract. Mahler equations relate evaluations of the same function \( f \) at iterated \( b \)th powers of the variable. They arise in particular in the study of automatic sequences and in the complexity analysis of divide-and-conquer algorithms. Recently, the problem of solving Mahler equations in closed form has occurred in connection with number-theoretic questions. A difficulty in the manipulation of Mahler equations is the exponential blow-up of degrees when applying a Mahler operator to a polynomial. In this work, we present algorithms for solving linear Mahler equations for series, polynomials, and rational functions, and get polynomial-time complexity under a mild assumption. Incidentally, we develop an algorithm for computing the \( \gcd \) of a family of linear Mahler operators with nonzero constant terms.

1. Introduction

1.1. Context. Our interest in the present work is in computing various classes of solutions to linear Mahler equations of the form

\[
\ell_r(x)y(x^{b^r}) + \cdots + \ell_1(x)y(x^b) + \ell_0(x)y(x) = 0,
\]

where \( \ell_0, \ldots, \ell_r \) are given polynomials, \( r > 0 \) is the order of the equation, and \( b \geq 2 \) is a fixed integer.

Mahler equations were first studied by Mahler himself in a nonlinear context [12]. His aim was to develop a general method to prove the transcendence of values of certain functions. Roughly speaking, the algebraic relations over \( \bar{\mathbb{Q}} \) between certain of these values come from algebraic relations over \( \bar{\mathbb{Q}}(z) \) between the functions themselves. This direction was continued by several authors. We refer to Pellarin’s introduction [14] for a historical and tutorial presentation, and to the references therein; see also Nishioka [13] for a textbook.

Mahler equations are closely linked with automata theory: the generating series of any \( b \)-automatic sequence is a Mahler function, that is, a solution of a linear Mahler equation; see [6, 5]. Mahler functions also appear in many areas at the interface of mathematics and computer science, including combinatorics of partitions, enumeration of words, and the analysis of divide-and-conquer algorithms.

Very recently, functional relations between Mahler functions have been further studied with a bias to effective tests and procedures [2, 7, 15]. Such studies motivate the need for algorithms that solve Mahler equations in various classes of functions. For instance, testing transcendence by the criterion of Bell and Coons [2] requires to date: December 19, 2016.

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compute a truncation of a Mahler series to suitable order, and the hypertranscendence
criterion by Dreyfus, Hardouin, and Roques [7] relies on determining if certain Mahler
equations possess ramified rational solutions.

1.2. Related work. Mahler equations are a particular case of difference equations, as can be seen by considering the change of variables $t = \log_b \log_b x$. Algorithms dealing with difference equations have been widely studied. For instance, the computation of rational solutions of linear difference equations with polynomial coefficients is an important basic brick coming up repeatedly in other algorithms. The reduction from Mahler to difference equation, however, does not preserve polynomial coefficients, which means that the above algorithms cannot be used in this setting.

There has been comparatively little interest in algorithmic aspects of Mahler equations. To the best of our knowledge, the only systematic study is by Dumas in his PhD thesis [8]. In particular, he describes procedures for computing various types of solutions of linear Mahler equations [8, Chapter 3]. However, beside a few gaps of effectiveness, that work does not take computational complexity issues into account. To a large extent, the results of the present work can be viewed as refinements of it, with a focus on efficiency and complexity analysis. More recently, Bell and Coons [2] give degree bounds that readily translate into algorithms for polynomial and rational solutions based on undetermined coefficients. With regard to series solutions, van der Hoeven [17, §4.5.3] suggests an algorithm that applies, under hypotheses, to certain equations of the form (EQN) as well as to certain nonlinear generalizations, and computes the first $n$ terms of a power series solution in $\tilde{O}(n)$ arithmetic operations. At least in the linear case and in analogy to the case of difference equations, this leaves the open question of an algorithm in complexity $O(n)$.

1.3. Setting. Our goal in this article is to present algorithms that compute complete sets of polynomial solutions, rational function solutions, truncated power series solutions, and truncated Puiseux series solutions of (EQN). More precisely, let $K$ be a (computable) subfield of $\mathbb{C}$, and suppose $\ell_0, \ldots, \ell_r \in K[x]$. Denote by $K((x^{1/n}))$ the field $\bigcup_{n=1}^{+\infty} K((x^{1/n}))$ of formal Puiseux series with coefficients in $K$. Let $M$ denote the Mahler operator of radix $b$, that is the automorphism of $K((x^{1/n}))$ that substitutes $x^b$ for $x$ and reduces to the identity map on $K$. Writing $x$ again for the operator of multiplication of a series by $x$, $M$ and $x$ follow the commutation rule $Mx = x^b M$. Equation (EQN) then rewrites as $Ly = 0$ where

$$(\text{opr}) \quad L = \ell_r M^r + \cdots + \ell_0$$

in the algebra generated by $M$ and $x$. We are interested in the algebraic complexity of computing the kernel of $L$ in each of $K[x], K(x), K[[x]]$, and $K((x^{1/n}))$.

We always assume that $\ell_r$ is nonzero. Except where otherwise noted, we also assume $\ell_0 \neq 0$. From a decidability viewpoint, the latter assumption is no loss of generality thanks to the following result [8, Cor. 6, p. 36].

**Proposition 1.1.** Given a linear Mahler equation of the form (EQN), one can compute an equation of the same form, with $\ell_0 \neq 0$, that has exactly the same formal Laurent series solutions—and therefore, the same polynomial solutions and the same rational-function solutions.

Note however that this result does not say anything about the cost of reducing to the case $\ell_0 \neq 0$. We give a complexity bound for this step in §4. As it turns
out, this bound often dominates our complexity estimates for the actual solving algorithms. Let us therefore stress that all other complexity results are stated under the assumption that $\ell_0$ is nonzero.

For $0 \leq k \leq r$, we denote by $v_k \in \mathbb{N} \cup \{+\infty\}$ and $d_k \in \mathbb{N} \cup \{-\infty\}$ the valuation and degree of the coefficient $\ell_k$. Let $d = \max_{0 \leq k \leq r} d_k$. Polynomials are implicitly represented in dense form, so that polynomials of degree $d$ in $\mathbb{K}[x]$ have size $d + 1$. All complexity estimates are given in terms of arithmetical operations in $\mathbb{K}$, which we denote “ops”. The complexity of multiplying two polynomials of degree at most $n$ is denoted by $M(n)$; we make the standard assumptions that $M(n) = O(n^2)$ and that $n \mapsto M(n)/n$ is nondecreasing.

Given two integers or polynomials $a$ and $b$, we denote their gcd by $a \land b$ and their lcm by $a \lor b$; we use $\wedge$ and $\lor$ for nary forms.

The following identities are used repeatedly in the text. We gather and repeat them here for easier reference:

\begin{align*}
\text{(EQN)} & \quad \ell_r(x)y(x^{b^r}) + \cdots + \ell_1(x)y(x^b) + \ell_0(x)y(x) = 0, \\
\text{(OPR)} & \quad L = \ell_rM^{b^r} + \cdots + \ell_0, \\
\text{(MU-NU)} & \quad \nu = \max_{k \geq 1} \frac{v_0 - v_k}{b^k - 1}, \quad \mu = v_0 + \nu.
\end{align*}

1.4. General strategy and outline. The article is organized as follows. In §2, we develop algorithms to compute truncated series solutions of equations of the form (EQN). We start with an example that illustrates the structure of the solution space and some of the main ideas behind our algorithms (§2.1). Then, we introduce a notion of Newton polygons, and use it to prove that the possible valuations (resp. degrees) of the solutions of (EQN) in $\mathbb{K}((x^{1/\nu}))$ (resp. $\mathbb{K}[x]$) belong to a finite set that we make explicit (§2.2). We compute a suitable number of initial coefficients by solving a linear system (§2.4), then prove that the following ones can be obtained iteratively in linear time, and apply these results to give a procedure that computes a complete set of truncated series solutions (§2.5). Finally, we extend the same ideas to the case of solutions in $\mathbb{K}[x]$ (§2.6) and in $\mathbb{K}((x^{1/\nu}))$ (§2.7).

The next section, §3, deals with solutions in $\mathbb{K}(x)$. The general idea is to first obtain a denominator bound, that is a polynomial $q$ such that $Lu = 0$ with $u \in \mathbb{K}(x)$ implies $qu \in \mathbb{K}[x]$ (§3.1). Based on elementary properties of the action of $M$ on elements of $\mathbb{K}[x]$ (§3.2), we give several algorithms for computing such bounds (§3.3–§3.4). This reduces the problem to computing a set of polynomial solutions with certain degree constraints, which can be solved efficiently using the primitives

\begin{table}[h]
\begin{tabular}{|l|l|l|}
\hline
Kind of solutions & Algorithm & Complexity \\
\hline
$\mathbb{K}[[x]]$, to order $|\nu| + 1$ & Alg. 4 & $O(rd\nu_0 + r^2M(v_0))$
$\mathbb{K}[[x]]$, to order $n$, when $r = O(d)$ & Alg. 3 & $O(rd^3 + nrd)$
$\mathbb{K}[x]$ & Alg. 6 & $O(b^{-r}d^2 + M(d))$
$\mathbb{K}((x^{1/\nu}))$ & Alg. 7 & $O(r^2Nd(d + n))$
$\mathbb{K}((x^{1/\nu}))$ & Alg. 7 & $O(r^2b^r(d + n))$
$\mathbb{K}(x)$, when $b = 2$ & Alg. 9 & $O(d^3)$
$\mathbb{K}(x)$, when $b \geq 3$ & Alg. 9 & $O(b^{-r}d^2)$
\hline
\end{tabular}
\caption{Complexity of the solving algorithms presented in the paper, assuming $\ell_0 \neq 0$.}
\end{table}
developed in§2. This leads to an algorithm for solving linear Mahler equations in $\mathbb{K}(x)$ (§3.5).

Finally, in §4, we generalize our study to the situation where the coefficient $\ell_0$ in (EQN) is zero. This makes us develop an unexpected algorithm for computing the gcrd of a family of operators, which we analyze and compare to the more traditional approach via Sylvester matrices and subresultants.

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2. Polynomial and series solutions

2.1. A worked example. The aim of this section is to illustrate our solving strategy in $\mathbb{K}[x]$ and $\mathbb{K}((x^{1/9}))$ on an example that we treat straightforwardly.

In radix $b = 3$, consider the equation $L y = 0$ where

$$L = x^3(1 - x^3 + x^6)(1 - x^7 - x^{10}) M^2 - (1 - x^{28} - x^{31} - x^{37} - x^{40}) M + x^6(1 + x)(1 - x^{21} - x^{30}).$$

Assume that $y \in \mathbb{K}((x^{1/9}))$ is a solution whose valuation is a rational number $v$. The valuations of $\ell_k M^k y$, for $k = 0, 1, 2$, are respectively equal to $6 + v, 3v, 3 + 9v$. If one of these rational numbers was less than the other two, then the valuation of the sum $\sum_{k=0}^2 \ell_k M^k y$ would be this smaller number, and $L y$ could not be zero. Consequently, at least two of the three rational numbers $6 + v, 3v, 3 + 9v$ have to be equal to their minimum. After solving, we find $v \in \{-1/2, 3\}$.

First consider the case $v = 3$, and write $y = \sum_{n \geq 3} y_n x^n$. For $m$ from 10 to 15, extracting the coefficients of $x^m$ from both sides of $0 = \ell_0 y + \ell_1 M y + \ell_2 M^2 y$, we find that $y_3, \ldots, y_9$ satisfy

$$0 = y_3 + y_4, \quad 0 = y_4 + y_5, \quad 0 = -y_4 + y_5 + y_6, \quad 0 = y_6 + y_7, \quad 0 = y_7 + y_8, \quad 0 = y_8 + y_9. \quad (2.2)$$

More generally, extracting the coefficient of $x^m$ yields the relation

$$y_{m+6} + y_{m+7} - y_{m-27} - y_{m-28} - y_{m-36} - y_{m-37} - (y_{m+3} - y_{m+28} - y_{m+31} - y_{m+37} - y_{m+40}) + (y_{m+3} - y_{m+6} - y_{m+9} - y_{m+10} - y_{m+19}) = 0, \quad (2.3)$$

where $y_s$ is understood to be zero if the rational number $s$ is not a nonnegative integer. This equation takes different forms, depending on the residue of $m$ modulo 9: for example, for $m = 20$ and $m = 42$, it reduces to, respectively,

$$y_{14} + y_{13} = 0, \quad y_{36} + y_{35} - y_{15} - 2y_{14} - y_{6} - y_{5} - y_{4} = 0.$$

Despite these variations, for any $m \geq 10$ the index $n = m - 6$ is the largest integer index occurring in (2.3). It follows that for successive $m \geq 10$, we can iteratively obtain $y_n$ from (2.3) in terms of already known coefficients of the series. Conversely, any sequence $(y_n)_{n \geq 3}$ that satisfies (2.3) gives a solution $y = \sum_{n \geq 3} y_n x^n$ of (EQN).
As a consequence, the power series solution is entirely determined by the choice of $\tilde{y}_3$ and the space of solutions of (EQN) in $K[[x]]$ has dimension one. A basis consists of the single series

\[(2.4) \quad x^3 - x^4 + x^5 - 2x^6 + 2x^7 - 2x^8 + 3x^9 - 3x^{10} + 3x^{11} - 5x^{12} + \ldots.\]

The other possible valuation, $v = -1/2$, is not a natural number. To revert to the simpler situation of the previous case, we perform the change of variables $x = t^2$ followed by the change of unknowns $y(t) = \tilde{y}(t)/t$. The equation becomes $\tilde{L}\tilde{y} = 0$ with

\[\begin{align*}
(2.5) \quad \tilde{L} &= (1 - t^6 + t^{12})(1 - t^{14} - t^{20})M^2 \\
&\quad - (1 - t^{56} - t^{62} - t^{74} - t^{80})M + t^{14}(1 + t^2)(1 - t^{42} - t^{60}).
\end{align*}\]

To understand this calculation, remember that $M$ was defined on $K((x^{1/2}))$, so that $M(t) = M(x^{1/2}) = x^{3/2} = t^3$.

We now expect $\tilde{L}$ to have solutions $\tilde{y} = \sum_{n \geq 0} \tilde{y}_n t^n$ of valuation 0 and 7 with respect to $t$, and the solutions of $\tilde{L}$ with valuation 0 to correspond to the solutions of $L$ with valuation $-1/2$. Extracting the coefficients of $x^m$ for $m$ from 0 to 24 from both sides of $\tilde{L}\tilde{y} = 0$ and skipping tautologies, we find that $\tilde{y}_0, \ldots, \tilde{y}_{10}$ satisfy

\[
\begin{align*}
0 &= -\tilde{y}_1, \\
0 &= -\tilde{y}_0 - \tilde{y}_2, \\
0 &= \tilde{y}_1 - \tilde{y}_3, \\
0 &= \tilde{y}_0 - \tilde{y}_4, \\
0 &= \tilde{y}_1 + \tilde{y}_2, \\
0 &= 2\tilde{y}_2 + \tilde{y}_4 - \tilde{y}_6, \\
0 &= \tilde{y}_3 + \tilde{y}_5 + \tilde{y}_6, \\
0 &= \tilde{y}_1 + \tilde{y}_3 + \tilde{y}_5, \\
0 &= -\tilde{y}_1 + \tilde{y}_7 + \tilde{y}_9, \\
0 &= -\tilde{y}_2 + \tilde{y}_8, \\
0 &= -\tilde{y}_2 + \tilde{y}_{10}.
\end{align*}
\]

Reasoning as above, we derive that, given $\tilde{y}_0$ while enforcing $\tilde{y}_7 = 0$, there is exactly one power series solution to $\tilde{L}$. More specifically when $\tilde{y}_0 = 1$ and $\tilde{y}_7 = 0$, we find the series

$1 - t^2 + t^4 - t^6 + t^{10} + t^{12} + \ldots$.

Hence, there is a 2-dimensional solution space in $K((x^{1/2}))$ for the original equation (EQN), with a basis consisting of the power series (2.4) and the additional Puiseux series

$\begin{align*}
x^{-1/2} - x^{1/2} + x^{3/2} - x^{5/2} + x^{7/2} - x^{9/2} + x^{11/2} + \ldots.
\end{align*}$

2.2. Valuations and degrees. Let us assume that $y \in K((x^{1/2}))$ is a solution of (EQN), whose valuation is a rational number $v$. The valuation of the term $\ell_k M^k y$ is then $v_k + b^k v$. Among those expressions, at least two must be minimal to permit the left-hand side of (EQN) to be 0: therefore, there exist distinct indices $k_1, k_2$.
between 0 and $r$ such that
\begin{equation}
(2.6) \quad v_k + b^k_1 v = v_{k_2} + b^k_2 v = \min_{0 \leq k \leq r} v_k + b^k v.
\end{equation}

This necessary condition for $L y = 0$ can be interpreted using a Newton polygon analogous to that of algebraic equations [18, Sec. IV.3.2-3]: to each monomial $x^j M^k$ in $L$, we associate the point $(b^k, j)$ in the first quadrant of the Cartesian plane endowed with coordinates $U$ and $V$ (see Fig. 1). We call the collection of these points the Newton diagram of $L$, and the lower (resp. upper) boundary of its convex hull the lower (resp. upper) Newton polygon of $L$. That two integers $k_1, k_2$ satisfy (2.6) exactly means that $(b^k_1, v_k_1)$ and $(b^k_2, v_k_2)$ belong to an edge $E$ of slope $-v$ of the corresponding lower Newton polygon.

Given an edge $E$ as above, an arithmetic necessary condition holds in addition to the geometric one just mentioned: the coefficients of the monomials of $L$ associated to points of $E$ must add up to zero. We call an edge with this property admissible.

**Example 2.1.** The lower Newton polygon of the operator (2.1) appears in dashed lines in Figure 1. It contains two admissible edges, corresponding to the valuations 3 and $-1/2$.

We get the following criterion, already stated in [8, p. 51] with a slightly different proof.

**Lemma 2.2.** Let $L$ be defined as in (OPR). The valuation $v$ of any formal Puiseux series solution of (EQN) is the opposite of the slope of an admissible edge of the lower Newton polygon of $L$. It satisfies
\begin{equation}
-\frac{v_r}{b^{r-1}(b-1)} \leq v = -\frac{v_{k_1} - v_{k_2}}{b^{k_1} - b^{k_2}} \leq \frac{v_0}{b-1},
\end{equation}
where $(b^{k_1}, v_{k_1})$ and $(b^{k_2}, v_{k_2})$ are the endpoints of the implied edge.

**Proof.** The fact that $v$ is the opposite of a slope together with its explicit form follow from (2.6) and the discussion above. There remains to prove the upper and lower bounds. The leftmost edge of the lower Newton polygon of $L$ provides the largest valuation and its slope $(v_k - v_0)/(b^k - 1)$ for some $k \geq 1$ is bounded below by $-v_0/(b-1)$. In the same way, the rightmost edge provides the smallest valuation and its slope, of the form $(v_r - v_k)/(b^r - b^k)$ for some $k < r$, is bounded above by $v_r/(b^r - b^{r-1})$. \qed

**Proposition 2.3.** The dimension of the space of solutions of the homogeneous equation $L y = 0$ in $\mathbb{K}((x^{1/s}))$ is bounded by the order $r$ of $L$.

**Proof.** The space of solutions admits a basis consisting of Puiseux series with pairwise distinct valuations. The number of possible valuations is bounded by the edge count of the lower Newton polygon of $L$, which is at most $r$. \qed

**Remark 2.4.** As we will see, the dimension of the solutions in $\mathbb{K}((x^{1/s}))$ can be strictly less than $r$. It is natural to ask how to construct a “full” system of $r$ linearly independent formal solutions in some larger extension of $\mathbb{K}(x)$. We will not pursue this question here and point to Roques’s work for an answer; see [15, Lemma 20 and Thm 35] and [16, Theorem 1]. See also Remark 2.18 below.
In analogy with the previous discussion on valuations of solutions, if a Puiseux series solution of (eqn) involves monomials with maximal exponent $\delta$, then the expression $d_k + b^k\delta$ must reach its maximum at least twice as $k$ ranges from 0 to $r$. As we see by the same reasoning as above (or by changing $x$ to $1/x$, which exchanges the lower and upper Newton polygons), $-\delta$ is then one of the slopes of the upper Newton polygon of $L$. The largest possible value corresponds to the rightmost edge.

**Lemma 2.5.** The maximum exponent $\delta$ of a monomial in a finite Puiseux series solution, and in particular the degree of a polynomial solution, is the opposite of the slope of an admissible edge of the upper Newton polygon. It satisfies

$$\delta = -\frac{d_{k_1} - d_{k_2}}{b^{k_1} - b^{k_2}} \leq \frac{d}{b^{r-1}(b-1)},$$

for some $k_1 \neq k_2$.

The admissibility of an edge of the upper Newton polygon is defined in analogy with admissibility in the lower Newton polygon.

2.3. **The nonhomogeneous case.** One of the proofs of results about Puiseux series solutions in §2.7 makes use of extended Newton diagrams that take into account the right-hand side of nonhomogeneous equations.
For $L$ as in (OPR) and a Puiseux series $\ell_{-\infty}$ of valuation $v_{-\infty} \in \mathbb{Q} \cup \{+\infty\}$, consider the nonhomogeneous equation

\begin{equation}
\ell_r(x)y(x^{b_r}) + \cdots + \ell_1(x)y(x^{b_1}) + \ell_0(x)y(x) = \ell_{-\infty}(x).
\end{equation}

Given a Puiseux series solution $y \in \mathbb{K}((x^{1/\alpha}))$ of this equation, with valuation $v \in \mathbb{Q}$, we define the Newton diagram of $(L, \ell_{-\infty})$ as the Newton diagram of $L$, augmented with all points $(0, \alpha)$ for which $x^{\alpha}$ appears with nonzero coefficient in $\ell_{-\infty}$. The notion of lower Newton polygon extends correspondingly.

As in §2.2, these definitions are motivated by analyzing the minimum of the valuations $v_k + b^kv$ of the terms of the left-hand side of (2.7): either this minimum is equal to $v_{-\infty}$, or it is less than $v_{-\infty}$ and must be reached at least twice on the left-hand side. In both cases, making the convention that $b^{-\infty} = 0$, there exist distinct indices $k_1, k_2$, now in $\{-\infty, 0, 1, \ldots, r\}$, such that the analogue

\[ v_{k_1} + b^{k_1}v = v_{k_2} + b^{k_2}v = \min_{k \in \{-\infty, 0, 1, \ldots, r\}} v_k + b^kv \]

of (2.6) holds. Again, this exactly means that $(\hat{b}^{k_1}, v_{k_1})$ and $(\hat{b}^{k_2}, v_{k_2})$ belong to an edge $E$ of slope $-v$ of the lower Newton polygon, now of $(L, \ell_{-\infty})$.

Depending on $v_{-\infty}$ and $\hat{v} = \min_{0 \leq k \leq r}(v_k + b^kv)$, the lower Newton polygon of $(L, \ell_{-\infty})$ can: be equal to that of $L$, if $\ell_{-\infty} = 0$; add an edge to its left,
Admissibility is again a necessary condition for

Assume that

Proposition 2.6.

equations

hence, for any given
determines

the row

the system

Proof.

then unique.

of

y

coefficients of a formal power series solution

v

the

solution when

ν <

as we can deduce from the proof of Lemma 2.2, there is no nonzero power series

and the

In terms of the Newton diagram,

more precisely in Proposition 2.6 below, define

determines a new coefficient

Because of our assumption that

nonzero, except maybe at intersection points of such lines (obtained for different

Y

row and column indices, starting at 0, as continuous variables

series solutions

Approximate series solutions.

2.4. Approximate series solutions. We now concentrate on the search for power

series solutions

where

of

k

is a vector of initial

coefficients of a formal power series solution

ν ≥ 0. A vector

is a vector of initial

coefficients of a formal power series solution

y = y_0 + \cdots + y_{\nu} x^{\nu} + y_{\nu+1} x^{\nu+1} + \cdots

of (EQN) if and only if it satisfies the linear system given by the upper left

submatrix of R. The power series solution (2.8) extending

is then unique.

Proof. A series

is a solution if and only if its coefficients satisfy

the system

(2.9)

v_0 + n < v_1 + b^1 n, \ldots, \quad v_0 + n < v_r + b^r n,

the row

of

is the first one with a nonzero entry of index

. It then determines

in terms of

Condition (2.9) is equivalent to

n > \nu,

hence, for any given

there is a unique choice of

satisfying all the equations

for

m > v_0 + \nu

As, when

holds for a given

the entries of
index \( n \) of \( R_m \) with \( m < v_0 + n \) are zero, the remaining equations \((R_m)_{0 \leq m \leq \mu}\) only involve the unknowns \((y_n)_{0 \leq n \leq \nu}\). \qed

We note in passing the following corollary, which is the essential argument in the proof of [15, Theorem 22].

**Corollary 2.7.** In case the leftmost edge of the lower Newton polygon of \( L \) lies on the axis of abscissas and is admissible, Equation (eqn) admits a power series solution of valuation 0.

**Proof.** We then have \( \nu = \mu = 0 \), so the only condition to check is that the first entry of \( R_0 \) is zero. This is equivalent to the edge being admissible. \qed

The geometric interpretation of the quantities \( \mu \) and \( \nu \) defined by (mu-nu) is a special case of a general correspondence between the structure of the matrix \( R \) and the Newton diagram of \( L \) via the point-line duality of plane projective geometry. The correspondence stems from the fact that a monomial \( x^jM^k \) of \( L \) is associated both to a point \((b^k,j)\) in the Newton diagram and, by considering its action on powers of \( x \), to the entries of \( R \) lying on the line \( Y = b^kX + j \). More generally, under projective duality, each point \((U,V)\) in the plane of the Newton diagram corresponds to a line \( Y = UX + V \) in the plane of the matrix \( R \), while, conversely, the dual of a point \((X,Y)\) is the line \( V = -UX + Y \). A line through two points \((U_1,V_1)\) and \((U_2,V_2)\) corresponds to the intersection of their duals.

In particular, the point \( P_0 = (1,v_0) \) corresponds to the right boundary \( \Delta : Y = X + v_0 \) of the strip of entries of slope 1 in the matrix \( R \) (see Figures 1 and 2). In the \((U,V)\)-plane, the line containing the leftmost edge of the lower Newton polygon passes through that point \( P_0 = \Delta^* \). This line is \( \Lambda : V = -\nu U + \mu \) and corresponds to the bottommost intersection \( \Lambda^* = (\nu,\mu) \) of \( \Delta \) with the right boundary of another strip. Below this intersection, the entries of \( R \) lying on \( \Delta \) are the topmost nonzero entries of their respective columns, and, at the same time, the rightmost nonzero entries of their respective rows: as already observed, each row \( R_m \) then determines a new \( y_n \).

**Example 2.8.** For the operator \( L \) of §2.1, the right boundaries of the strips associated to the three terms of \( L \) have equations \( Y = X + 6 \), \( Y = 3X \), and \( Y = 9X + 3 \) respectively (dotted lines in Fig. 2). The first two of them meet at \( \Lambda^* = (3,9) \) (Fig. 2, hollow circle at the bottom right corner of the gray rectangle), and the line \( \Delta : Y = X + 6 \) becomes the rightmost line for \( Y > 9 \). For \( m \geq 10 \), the row \( R_m \) reflects the relation (2.3). In particular, the existence of a power series solution is entirely determined by the small linear system that uses the rows \( R_0 \) to \( R_9 \) and the unknowns \( y_0 \) to \( y_3 \) (gray rectangle on Figure 2). Solving the system yields \( y_0 = y_1 = y_2 = 0 \) and \( y_3 \) arbitrary. We then recover the results of §2.1: the space of solutions of (eqn) in \( K[[x]] \) has dimension one and a basis consists of the single series (2.4). The \( V \)-intercept of the leftmost edge of the lower Newton polygon is \( \mu = 9 \), and the corresponding slope is \( -\nu = -3 \). In this case, it is both the column dimension of the small system and the valuation of the solution. Observe how the bottom right sector depicted in light gray corresponds to the system starting with equations (2.2): as the top left rectangle imposes \( y_0 = y_1 = y_2 = 0 \), the dots on the left of the sector in light gray play no role in the equations.

As we will see, in the situation of Proposition 2.6, the coefficients \( y_{[\nu]+1} \) to \( y_{[\nu]+\kappa} \) of \( y \) can be computed from \( y_0, \ldots, y_{[\nu]} \) in \( O(n) \) ops for fixed \( L \). This motivates to
Input: The linear Mahler equation \((\text{EQN})\). A transformation \(\phi\) of the form \((2.10)\). Integers \(w,h\). A set \(E = \{m_0, m_1, \ldots\}\) of row indices, with \(m_0 < m_1 < \cdots < h\).

Output: The submatrix \(R_E = (R_{m,n})_{0 \leq n < w}\) of the infinite matrix \(R(\phi(L))\).

1. Initialize a row-sparse \(|E| \times w\) matrix \(R_E\).
2. For \(i = 0, 1, \ldots, |E| - 1\) and \(k = 0, 1, \ldots, r:\)
   (a) Set \(B = m_i + \gamma - \alpha b^k\).
   (b) Compute \(j_0' = \beta^{-1} B \mod b^k\) (with \(0 \leq j_0' < b^k\)).
   (c) For \(j' = j_0', j_0' + b^k, j_0' + 2b^k, \ldots\) while \(j' \leq d_k\) and \(\beta j' \leq B:\)
      (i) If \(\beta j' > B - b^k w\), then add \(\ell_{k,j'}\) to \(A[i, b^k(B - \beta j')]\).
3. Return \(R_E\).

Algorithm 1. Matrix \(R\).

call the truncation to order \(O(x^{\lceil \nu \rceil + 1})\) of a series solution an approximate series solution of \((\text{EQN})\).

2.5. Power series solutions. Our goal at this point is to describe an algorithm that computes the formal power series solutions of \((\text{EQN})\), truncated to any specified order. We first explain how to compute the entries of the matrix \(R\). It is convenient, for expository reasons, to frame this computation as an individual step that returns a sparse representation of a submatrix of \(R\) corresponding to a subset of the rows. Indeed, in our complexity model dense matrices could not lead to good bounds. We therefore define a matrix representation to be row-sparse if iterating over the nonzero entries of any given row does not require any zero test in \(K\). Then, the algorithm essentially amounts to an explicit expression for the coefficients of recurrences similar \((2.3)\), which can as well be computed on the fly.

In view of the computation of ramified solutions (§2.7), Algorithms 1 and 2 accept as input a \(K\)-linear transformation \(\phi\) to be applied to the operator \(L\). In general, \(\phi\) will take the form

\[(2.10) \quad \phi(x^j M^k) = x^{\alpha b^k + \beta j - \gamma} M^k, \quad \alpha, \gamma \in \mathbb{Z}, \quad \beta \in \mathbb{N}_{>0}, \quad \beta \lambda b = 1,
\]

with \(\alpha, \beta, \gamma\) chosen such that \(\phi(L)\) has plain (as opposed to Laurent) polynomial coefficients. The reader only interested in polynomial, rational, and power series solutions of \(L\) may safely assume \(\phi = \text{id}\), i.e., \(\alpha = \gamma = 0, \beta = 1\).

Lemma 2.9. Algorithm 1 computes the submatrix \(R_E\) obtained by taking the first \(w\) entries of the rows of \(R(\phi(L))\) with index \(m \in E\) in \(O((r + d)|E|)\) ops. Each row of \(R_E\) has at most \(r + 2d\) nonzero entries.

Proof. Write \(\tilde{L} = \sum_{k=0}^{r} \tilde{\ell}_k(x) M^k = \phi(L)\). Recall that the row \(R_m\) is obtained by extracting the coefficient of \(x^m\) in the equality \(\tilde{L}y = 0\), where \(y = \sum_{n \geq 0} y_n x^n\). More precisely, \(R_{m,n}\) is the coefficient of \(y_n x^m\) in the series

\[
\tilde{L}y = \sum_{k=0}^{r} \sum_{j=0}^{d} \tilde{\ell}_{k,j} x^j \sum_{n=0}^{\infty} y_n x^{k+n} = \sum_{m=0}^{\infty} \sum_{n=0}^{\infty} \left( \sum_{j+b^k n=m} \tilde{\ell}_{k,j} \right) y_n x^m.
\]
**Input:** A linear Mahler operator $L$ of order $r$. A transformation $\phi$ of the form (2.10). Integers $h, w \in \mathbb{N}$. A set $E = \{m_0, \ldots, m_{w-1}\}$ with $m_0 < \cdots < m_{w-1} < h$, such that the submatrix $(R_{m,j})_{0 \leq i,j < w}$ of $R(\phi(L))$ is lower, resp. upper, triangular, with at most $r$ zeros on the diagonal.

**Output:** A vector $(f_1, \ldots, f_r)$ of polynomials of degree less than $w$.

(1) Construct the row-sparse submatrix $S_E = (R_{m,i,j})_{0 \leq i,j < w}$ by Algorithm 1.

(2) Compute a basis of $\ker S_E$ as a matrix $G = (G_{i,j}) \in \mathbb{K}^{w \times \rho}$ by forward, resp. backward, substitution, using the row-sparse structure.

(3) For $1 \leq j \leq \rho$, set $g_j = G_{0,j} + G_{1,j}x + \cdots + G_{w-1,j}x^{w-1} \in \mathbb{K}[x]$ and compute the coefficients of $Lg_j(x) \mod x^h = \sum_{0 \leq i < h} s'_{i,j}x^i$, then form the matrix $S' = (s'_{i,j}) \in \mathbb{K}^{h \times \rho}$.

(4) Compute a basis of $\ker S'$ as a matrix $K \in \mathbb{K}^{\rho \times \sigma}$ by the algorithm of Ibarra, Moran and Hui [11].

(5) Compute $F = (F_{i,j}) = GK \in \mathbb{K}^{w \times \sigma}$.

(6) Return $(f_1, \ldots, f_r)$ where $f_j = F_{0,j} + \cdots + F_{w-1,j}x^{w-1}$.

Algorithm 2. Solutions over prescribed monomial support.

The definition of $\phi$ translates into $\ell_{k,j} = 0$ when $j \not\equiv ab^k - \gamma \pmod{\beta}$, and otherwise $\ell_{k,j} = \ell_{k,j'}$ for $j \equiv ab^k + \beta j' - \gamma$. Therefore, $R_{m,n}$ is equal to the sum of $\ell_{k,j'}$ for $(k, j')$ satisfying $ab^k + \beta j' - \gamma = m - nb^k$. For fixed $m$ and $k$, the coefficient $\ell_{k,j'}$ only contributes when $\beta j' \equiv m + \gamma \pmod{b^k}$. Its contribution is then to $R_{m,n}$ with $n = b^{-k}(B - \beta j')$ where $B = m + \gamma - ab^k$, and we are only interested in $0 \leq n < w$, i.e., $B - b^kw < \beta j' \leq B$. Using the assumption that $\beta$ is coprime with $b$, the condition on $\beta j' \pmod{b^k}$ rewrites as $j' \equiv j_0' \pmod{b^k}$, where $j_0'$ is the integer computed at step 2b. Therefore, the loop 2c correctly computes the contribution of $\ell_k$ to the entries of index less than $w$ of the row $R_{m,i}$, and hence the algorithm works as stated.

The only operations in $\mathbb{K}$ performed by the algorithm are one addition and possibly one comparison (to update the sparse structure) at each loop pass over step 2(c)i. The total number of iterations of the innermost loop for a given $i$ is at most

\[
\sum_{k=0}^{r} \left\lfloor \frac{d_k}{b^k} \right\rfloor \leq r + \frac{b}{b-1}d \leq r + 2d
\]

and bounds the number of nonzero entries in the row of index $m_i$. The complexity in ops follows by summing over $i$. \hfill \Box

According to Proposition 2.6, the number of linearly independent power series solutions and their valuations are determined by a small upper upper left submatrix of $R$. As a direct attempt at solving the corresponding linear system would have too high a complexity (see Remark 2.11), our approach is to first find a set of candidate solutions, spanning a low-dimensional vector space that contains the approximate series solutions, and to refine the solving in a second step. Geometrically, the
idea to obtain a candidate solution \( g = g_0 + g_1 x + \cdots \) is to follow the “profile” of \( R \) (more precisely, the right boundary of the overlapping strips described in the previous section), using a single equation \( R_m \) to try and compute each coefficient \( g_n \) from \( g_0, \ldots, g_{n-1} \). (That is, for each \( n \), we resolutely skip all but one equations susceptible to determine \( g_n \).) By duality, this corresponds to keeping a varying line of increasing integer slope in contact with the lower Newton polygon, and having it “pivot” around it. In this process, the only case that potentially leaves a degree of freedom in the choice of \( g_n \) is when column \( n \) contains a “corner” of the profile, corresponding to an edge of the Newton polygon. As a consequence, it is enough to construct at most \( r \) independent candidates solutions. The second step then consists in recombining the candidates in such a way that the equations \( R_m \) that were skipped in the first phase be satisfied.

This strategy is made more precise in Algorithm 2, which will then be specialized to power series solutions (and later to other types of solutions) by a suitable choice of \( E, h \) and \( w \). By construction, Algorithm 2 outputs polynomials of degree less than \( w \) that are solutions of a subsystem of the linear system induced by \( L \). These polynomials need not a priori prolong into actual solutions.

**Lemma 2.10.** Algorithm 2 runs in \( O(rw d + r^2 w + r^2 M(h)) \) ops, and returns a basis of the kernel of the linear map induced by \( \phi(L) \) from \( \mathbb{K}[x]_{< w} \) to \( \mathbb{K}[x]/(x^h) \).

**Proof.** When \( S_E \) is lower, respectively upper, triangular it is possible at step 2 to compute \( G \) by forward, respectively backward, substitution, in such a way that \( S_E G = 0 \). By interpreting the \( h \times w \) upper left submatrix \( S \) of \( R \) as the matrix of a restriction of \( L \) to suitable monomial bases, it follows from the definition of \( S' \) that \( S' = S G \). Step 4 computes \( K \) such that \( S' K = 0 \).

The columns of \( F \), computed as \( G K \) at step 5, span the kernel of \( S \): Indeed, assume \( S f = 0 \), so that by selecting rows \( S_E f = 0 \), and \( f \) can be written as \( G \gamma \) for some \( \gamma \). Then, \( S' \gamma = S G \gamma = S f = 0 \). But this means that \( \gamma = K \eta \) for some \( \eta \), so that \( f = G K \eta = F \eta \). Conversely, we have \( S F = S G K = S' K = 0 \), so that any vector of the form \( F \eta \) belongs to \( \ker S \).

Additionally, since the columns of \( G \), respectively those of \( K \), are linearly independent, \( G K \eta = 0 \) implies \( K \eta = 0 \), which implies \( \eta = 0 \). The columns of \( F = G K \) hence form a basis of \( \ker S \).

By Lemma 2.9, step 1 takes \( O(w(\rho + d)) \) ops. The number of nonzero entries in each row of \( S_E \) is bounded by \( r + 2d \) by Lemma 2.9, hence the cost of computing \( \rho \) linearly independent solutions by substitution at step 2 is \( O(\rho w(\rho + d)) \). As no more than \( r \) of the diagonal entries of \( S_E \) are zero, \( \rho \) is at most \( r \). The computation of each column of \( S' \) at step 3 amounts to adding \( r + 1 \) products of the \( \ell_k \) by the \( M^k S_j \), truncated to order \( h \), for a total of \( O(r^2 M(h)) \) ops. As \( \rho \leq r \), computing the kernel of \( S' \) at step 4 via an LSP decomposition (a generalization of the LUP decomposition) requires \( O(h r^{\omega - 1}) = o(r^2 M(h)) \) ops [11]. Finally, the recombination at step 5 takes \( O(w r^{\omega - 1}) = o(r^2 w) \) ops as \( \sigma \leq \rho \leq r \).

**Remark 2.11.** Note that a direct attempt to solve \( S \), when, say, \( \phi = \text{id} \) and \( w = O(d) \), would result in a complexity \( O(d^2) \) (e.g., using the LSP decomposition), as opposed to \( O(d^2) \) when using Algorithm 2 and disregarding the dependency in \( r \).
We now specialize the generic solver to the computation of approximate series with (2.11). A polynomial \( \hat{y} = y_0 + \cdots + y_{|\hat{\nu}|} x^{|\hat{\nu}|} \) such that \( \phi(L) \hat{y} = O(x^{|\hat{\nu}|+1}) \), for \( \hat{\nu} \) and \( \hat{\mu} \) defined by (2.11). An integer \( n \).

### Proof. 

First of all, when \( m = m_i \in E \), none of the terms \( \tilde{\ell}_k M^k \) of \( \phi(L) \) contributes to the entries of \( S \) located above \( S_{m,n} \). The matrix \( S_E \) is thus lower triangular. In addition, \( R_{m,n} \) is zero (and only) if \( -n \) is an (admissible) slope of the lower Newton polygon, so that no more than \( r \) of the diagonal entries of \( S_E \) are zero. Both preconditions of Algorithm 2 are therefore satisfied. By Proposition 2.6 and Lemma 2.10, it follows from the choice of \( h \) and \( w \) that the \( f_j \) form a basis of approximate series solutions. Using the inequalities \( h \leq b_0/(b-1) + 1 = O(\tilde{v}_0) \) and \( w \leq v_0/(b-1) + 1 = O(\tilde{v}_0) \) in the formula of Lemma 2.10, the total complexity is as announced.

Given an approximate series solution, the next terms of the corresponding solutions can be computed efficiently one by one using simple recurrence formulae.

### Proposition 2.12. 

Assume \( \hat{\nu} \geq 0 \). Algorithm 2, called with 

\[
  h = |\hat{\mu}| + 1, \quad w = |\hat{\nu}| + 1, \quad E = \left( \min_k (\tilde{v}_k + nb^k) \right)_{0 \leq n < w},
\]

runs in \( O(rd\tilde{v}_0 + r^2 M(\tilde{v}_0)) \) ops and returns a basis of approximate series solutions of the equation \( \phi(L) y = 0 \).

### Proof. 

First of all, when \( m = m_i \in E \), none of the terms \( \tilde{\ell}_k M^k \) of \( \phi(L) \) contributes to the entries of \( S \) located above \( S_{m,n} \). The matrix \( S_E \) is thus lower triangular. In addition, \( R_{m,n} \) is zero (if and only if) only if \( -n \) is an (admissible) slope of the lower Newton polygon, so that no more than \( r \) of the diagonal entries of \( S_E \) are zero. Both preconditions of Algorithm 2 are therefore satisfied. By Proposition 2.6 and Lemma 2.10, it follows from the choice of \( h \) and \( w \) that the \( f_j \) form a basis of approximate series solutions. Using the inequalities \( h \leq b_0/(b-1) + 1 = O(\tilde{v}_0) \) and \( w \leq v_0/(b-1) + 1 = O(\tilde{v}_0) \) in the formula of Lemma 2.10, the total complexity is as announced.

Given an approximate series solution, the next terms of the corresponding series solutions can be computed efficiently one by one using simple recurrence formulae.

### Proposition 2.13. 

Given an approximate series solution \( \hat{y} = y_0 + \cdots + y_{|\hat{\nu}|} x^{|\hat{\nu}|} \) of (EQN), Algorithm 3 computes the truncation to the order \( O(x^{|\hat{\nu}|+n}) \) of the unique solution \( y \) of (EQN) of the form \( y = \hat{y} + O(x^{|\hat{\nu}|+1}) \) in \( O((r + d) n) \) ops.

### Proof. 

By Proposition 2.6, the system to be solved at step 2 is compatible. According to the description of \( R \) provided above, the submatrix \( (R_{m,n})_{m > |\hat{\nu}|, n > |\hat{\nu}|} \) is lower triangular, with nonzero diagonal coefficients, so that the system can be solved by forward substitution. As explained in §2.4, the output is a truncation of a solution.
Input: A linear Mahler operator \( L \) of order \( r \).
Output: A basis \( (f_1, \ldots, f_\sigma) \) of approximate series solutions of \( L \).

Let \( \mu, \nu \) be as defined by (MU-Nu). If \( \nu < 0 \), return \( () \). Otherwise, call Algorithm 2 with \( \phi = \text{id} \),
\[
h = \lfloor \mu \rfloor + 1, \quad w = \lceil \nu \rceil + 1, \quad E = \left( \min_{k} (v_k + nb^k) \right)_{0 \leq n < w},
\]
and return the result.

Algorithm 4. Approximate series solutions.

of \( \phi(L) \). By Lemma 2.9, the cost in ops of step 1 is \( O((r + d) n) \), and each row of \( S \) contains at most \( r + 2d \) nonzero entries. Therefore, step 2 costs \( O((r + d) n) \) ops.

2.6. Polynomial solutions. Our goal in this subsection is Algorithm 6, which computes a basis of all polynomial solutions. Lemma 2.5 provides us with an upper bound \( \frac{d}{b^r - b^{r-1}} + 1 = O(d/b^r) \) for the degree of any polynomial solution. Before we take this into account, we provide an algorithm to compute polynomial solutions with degree bounded by \( w \geq 0 \), which runs in a complexity that is sensitive to \( w \).

In the same way as in Proposition 2.12, to obtain candidate polynomial solutions \( f = f_0 + \cdots + f_w x^w \), we set \( f_n = 0 \) for \( n \geq w \) and then compute \( f_n \) for decreasing \( n \) by “following” the “left profile” of the matrix \( R \) (or, dually, the upper Newton polygon). The corresponding specialization of Algorithm 2 is formalized as Algorithm 5.

Proposition 2.14. Assume \( \nu \geq 0 \). Algorithm 2, called with \( \phi = \text{id} \) and
\[
(2.12) \quad h = d + (w - 1)b^r + 1, \quad E = \left( \max_{k} (d_k + nb^k) \right)_{0 \leq n \leq w},
\]
returns a basis of the space of polynomial solutions of (EQN) of degree less than \( w \). For \( w = O(d/b^r) \), the algorithm runs in \( \tilde{O}(wd + M(d)) \) ops.

Proof. The proof is similar to that of Proposition 2.12: the extracted submatrix of \( R \) is now upper triangular; the zeros on its diagonal correspond to the admissible nonpositive integer slopes of the upper Newton polygon; the number of such zeros is not more than \( r \). Both preconditions of Algorithm 2 are therefore satisfied and Lemma 2.10 applies. Additionally, the choice of \( h \) in terms of \( w \) is such that \( \deg(Ly) < h \) whenever \( \deg y < w \) for a polynomial \( y \). So, the basis returned is that of the kernel of the map induced by \( L \) from \( \mathbb{K}[x]_{<w} \) to \( \mathbb{K}[x] \), as announced.

For the complexity result, the hypothesis on \( h \) implies \( h = O(d) \) and \( r = O(\log_d d) \), so that the conclusion of Lemma 2.10 specializes to \( \tilde{O}(wd + M(d)) \) ops.

Remark 2.15. The loose bound on \( w \), namely \( w = O(d/b^r) \), permits in particular to obtain a result when \( d \) is not the maximal degree of the \( \ell_k \), but only bounds them up to a multiplicative constant. In this case, the complexity announced by Proposition 2.14 specializes to the same complexity as in Corollary 2.16. This will be used for the numerators of rational-function solutions in §3.5.
**Input:** A linear Mahler operator \( L \) of order \( r \). An integer \( w \in \mathbb{N} \).

**Output:** A basis \((f_1, \ldots, f_\sigma)\) of the polynomial solutions of \( L \) of degree less than \( w \).

Let \( \mu, \nu \) be as defined by \((\mu - \nu)\). If \( \nu < 0 \), return \( () \). Otherwise, call Algorithm 2 with \( \phi = \text{id} \), \( h = \max \left( \frac{d_k}{b^r - 1} \right) + 1 \), \( w = \max(d_k + nb^k)_{0 \leq n < w} \), and return the result.

**Algorithm 5.** Polynomial solutions of bounded degree.

**Input:** A linear Mahler operator \( L \) of order \( r \).

**Output:** A basis \((f_1, \ldots, f_\sigma)\) of all polynomial solutions of \( L \).

Call Algorithm 5 with \( w = \left\lfloor \frac{d}{b^r - 1} \right\rfloor + 1 \) and return the result.

**Algorithm 6.** Basis of polynomial solutions.

By Lemma 2.5, the degree of any polynomial solution is bounded above by \( \delta_0 = \frac{d}{b^r - b^{r-1}} + 1 \). Specializing Proposition 2.14 to \( w = \left\lfloor \delta_0 \right\rfloor \), we obtain a bound for the complexity of computing the whole space of polynomial solutions.

**Corollary 2.16.** Assuming \( \nu \geq 0 \), Algorithm 2, called with \( \phi = \text{id} \),

\[
h = 3d + 1, \quad w = \left\lfloor \frac{d}{b^r - 1} \right\rfloor + 1, \quad E = \left( \max(d_k + nb^k) \right)_{0 \leq n \leq w},
\]

computes a basis of the polynomial solutions of \((\text{EQN})\) in \( \bar{O}(d^2 / b^r + \text{M}(d)) \) ops.

**Proof.** Observe that the choice for \( w \) induces that \( h \), as defined in Algorithm 6, satisfies \( h \leq 3d + 1 \). The result follows from this fact and \( w = O(d / b^r) \).

2.7. **Puiseux series solutions.** We now discuss the computation of solutions of \((\text{EQN})\) in \( \mathbb{K}((x^{1/r})) \). Even though Proposition 1.1 does not apply, we still assume that the coefficient \( \ell_0 \) of \( L \) is nonzero. There is no loss of generality in doing so: if \( L = L_1 M^w \) for some \( w \in \mathbb{N} \), then the Puiseux series solutions of \( L \) are exactly the \( y(x^{b^{-w}}) \) where \( y \) ranges over the Puiseux series solutions of \( L_1 \). Additionally, the order of \( L_1 \) is bounded by that of \( L \), so that the complexity estimates depending on it will still hold (and equations of order zero that result from the transformation when \( r = w \) have no nontrivial solutions).

The computation of solutions \( y \in \mathbb{K}((x^{1/N})) \) with a given ramification index \( N \) is similar to that of power series solutions. In order to compute a full basis of solutions in \( \mathbb{K}((x^{1*/})) \), however, we need a bound on the ramification index necessary to express them all. Lemma 2.17, communicated to us by Dreyfus and Roques, and Proposition 2.19 below provide constraints on the possible ramification indices.

**Lemma 2.17.** If \( y \in \mathbb{K}((x^{1/r})) \) is a Puiseux series such that \( Ly \in \mathbb{K}((x^{1/q})) \) where \( q' \) is coprime with \( b \), then \( y \in \mathbb{K}((x^{1/q})) \) for some \( q \) coprime with \( b \).
Proof. Let $q_0$ be the smallest positive integer such that $y \in \mathbb{K}(x^{1/q_0})$. Set $g = q_0 \land b$ and $q'' = q_0/g$, so that $M_y \in \mathbb{K}(x^{b/q_0}) \subset \mathbb{K}(x^{1/q''})$. The expression
\[ y = \ell_0^{-1} \left( L_y - (\ell_1 + \cdots + \ell_r M_r^{-1}) M_y \right) \]
shows that $y \in \mathbb{K}(x^{1/q''})$ where $q_1 = q''$. By minimality of $q_0$, we have $q_1 = kq_0$ for some $k \in \mathbb{N}$, which simplifies to $q' = kg$. Since $q'$ was assumed to be coprime with $b$, this implies $g = 1$. \qed

Remark 2.18. Some non-Puiseux formal series solutions of Mahler equations with $\ell_0 \neq 0$ do involve ramifications of order divisible by $b$: perhaps the simplest example, akin to [4, p. 64] (see also [1]), is $y = x^1/b + x^1/b^2 + x^1/b^3 + \cdots$, which satisfies $\left(M - x^{b^{-1}}\right)(M - 1) y = 0$.

The following proposition formalizes, as a consequence of Lemma 2.17 and the properties of Newton polygons discussed in §2.2, that no ramification is needed beyond those present in the candidate leading terms given by the Newton polygon. Call $\mathcal{N}$ the lower Newton polygon of $L$, and let $Q$ denote the set of denominators $q$ of slopes (written in lowest terms) of admissible edges of $\mathcal{N}$ such that $q \land b = 1$.

Proposition 2.19. Any Puiseux-series solution $y$ of $L_y = 0$ belongs to $\mathcal{V} = \sum_{q \in Q} \mathbb{K}(x^{1/q})$. In particular, the space of solutions of $L$ in $\mathbb{K}(x^{1/N})$ is contained in $\mathbb{K}(x^{1/N})$, where $N \leq b^r - 1$ denotes the lcm of the elements of $Q$.

Proof. Let $y \in \mathbb{K}(x^{1/N})$ satisfy $L_y = 0$, and suppose by contradiction that $y$ contains a nonzero term of exponent $p_1/q_1$ where $p_1 \land q_1 = 1$ and $q_1$ does not divide any element of $Q$. Choose $p_1/q_1$ minimal with these properties. Write $y = y_0 + y_1$ where $y_0$ consists of the terms of $y$ with exponent strictly less than $p_1/q_1$, so that $y_0 \in \mathcal{V}$ and $y_1$ has valuation $p_1/q_1$. Then $g = L y_0$ belongs to $\mathcal{V}$, so that there exists $q'' \in \mathbb{N}$ for which $q'' \land b = 1$ and $g \in \mathbb{K}(x^{1/q''})$. Since $L y_1 = -g$, Lemma 2.17 implies that $y_1 \in \mathbb{K}(x^{1/q''})$ for some $q$ coprime with $b$. In particular, $q_1$ is coprime with $b$.

Since $p_1/q_1$ is the valuation of a solution of the equation $L z = -g$, its opposite $s = -p_1/q_1$ is the slope of an admissible edge $E$ of the lower Newton polygon $\mathcal{N}_g$ of $(L, -g)$ (see §2.3). On the other hand, because of the definition of $Q$ and the properties $q_1 \land b = 1$ and $q_1 \notin Q$, the edge $E$ cannot be an edge of $\mathcal{N}$. Therefore, by the description in §2.3, $g$ must be nonzero and the edge $E$ must be the leftmost edge of $\mathcal{N}_g$. The valuation of $g \in \mathcal{V}$ is thus a rational number $p_0/q_0$ (not necessarily in lowest terms) with $q_0 \in Q$, so that in particular $q_0 \land b = 1$. As $s$ is the slope of $E$ in $\mathcal{N}_g$, it is of the form $(q_0 v_k - p_0)/(q_0 b^k)$ for some $k \in \{0, \ldots, r\}$. Then, $q_1$ divides $q_0 b^k$. As it is coprime with $b$, this implies that $q_1$ divides $q_0 \in Q$, a contradiction. We have proved that $y$ belongs to $\mathcal{V}$.

Next, it is clear that $\mathcal{V}$ is contained in $\mathbb{K}(x^{1/N})$. Finally, letting $(b^k, v_i)$ denote the vertices of $\mathcal{N}$ (sorted from left to right as $i$ increases), the lcm $N$ satisfies $N \leq \prod_i (b^{k_{i+1} - k_i} - 1) < b^r$, as claimed. \qed

Remark 2.20. The bound $N < b^r$ is tight, as shown by the example of $M_r - x$, which admits the solution $x^{1/(b^r - 1)}$.

In order to obtain an algorithm that computes a basis of the space of Puiseux series solutions, there remains to generalize the results of §2.4–2.5 to the case of solutions lying in $\mathbb{K}(x^{1/N})$ where $N$ is given. Motivated by the structure of the space $\mathcal{V}$ described in Proposition 2.19, we do not require here that $N$ be equal to
the lcm of all elements of $Q$: setting it to the lcm of any subset of these elements also makes sense. For the most part, the algorithms searching for power series solutions apply \textit{mutatis mutandis} when the indices $m$ and $n$ are allowed to take negative and noninteger rational values. Nevertheless, some care is needed in the complexity analysis, so we explicitly describe a way to reduce the computation of ramified solutions of $L$ to that of power series solutions of an operator $\tilde{L}$.

Denote $x = t^\beta$, and consider the change of unknown functions $y(x) = t^\alpha z(t)$, for $\alpha \in \mathbb{Z}$ and $\beta \in \mathbb{N}_{>0}$ to be determined. Observe that $Mt = t^\beta$. If $y(x)$ is a solution of $Ly = 0$, then $z(t)$ is annihilated by

$$\tilde{L} = t^{-\gamma}L t^\alpha = t^{-\gamma} \sum_{k=0}^{r} t^{\alpha b_k} \ell_k(t^\beta)M^k = \sum_{k=0}^{r} \tilde{\ell}_k(t)M^k$$

where $\gamma \in \mathbb{Z}$ can be adjusted so that the $\tilde{\ell}_k$ belong to $K[t]$. We then have $\tilde{L} = \phi(L)$ where $\phi$ is the $K$-linear map, already introduced in \S2.5, that sends $x^j M^k$ to

$$\phi(x^j M^k) = t^{-\gamma} t^{\beta j} M^k t^\alpha = t^{-\gamma + \beta j + \alpha b_k} M^k.$$ 

Viewing monomials $x^j M^k$ as points in the plane of the Newton diagram, the map $\phi$ induces an affine shearing

$$[\phi]: \begin{pmatrix} b_k \\ j \end{pmatrix} \mapsto \begin{pmatrix} 1 & 0 \\ \alpha & \beta \end{pmatrix} \begin{pmatrix} b_k \\ j \end{pmatrix} + \begin{pmatrix} 0 \\ -\gamma \end{pmatrix}. $$

As in \S2.5, denote by $\tilde{v}_k$ and $\tilde{d}_k$ the valuations and degrees of the coefficients of $\tilde{L}$, and by $\tilde{\mu}$ and $\tilde{\nu}$ the quantities defined by (MU-NU) with $v_k$ replaced by $\tilde{v}_k$.

\textbf{Lemma 2.21.} Fix an edge $S_0$ of the lower Newton polygon of $L$, of slope $-p/q$ for (not necessarily coprime) $p \in \mathbb{Z}$ and $q \in \mathbb{N}$. Let $c$ be the $V$-intercept of the line supporting $S_0$. Set $\alpha = p$, $\beta = q$, and $\gamma = qc$ in (2.13). Then:

(a) the operator $\tilde{L} = \phi(L)$ has polynomial coefficients;
(b) its Newton diagram is the image of that of $\tilde{L}$ by $[\phi]$, with the edge $S_0$ being mapped to a segment of the $U$-axis;
(c) in terms of those of $L$, the parameters associated to $\tilde{L}$ satisfy

$$\tilde{d}_k = -qc + pb_k + qd_k \geq \tilde{v}_k = -qc + pb_k + qv_k \geq 0,$$

$$\tilde{\nu} = q\nu - p \geq 0, \quad \tilde{\mu} = q(\mu - c) \geq 0.$$
\textbf{Input}: A linear Mahler operator $L$ as in (OPT). A ramification index $N \in \mathbb{N}_{>0}$. A truncation order $n \in \mathbb{N}$.

\textbf{Output}: A vector $(\hat{y}_1, \ldots, \hat{y}_\sigma)$ of truncated Puiseux series.

1. Compute the slope $s$ and $V$-intercept $c$ of the rightmost admissible edge of the lower Newton polygon of $L$ with slope in $N^{-1}\mathbb{Z}$.
2. Define $\phi$ and $\tilde{L} = \phi(L)$ according to (2.13), with $\alpha = -Ns$, $\beta = N$, and $\gamma = Nc$.
3. Call Algorithm 2 on $L$ and $\phi$, with
   \[ h = \lceil \tilde{\mu} \rceil + 1, \quad w = \lceil \tilde{\nu} \rceil + 1, \quad E = (\min_k (\tilde{\nu}_k + nb^k))_{0 \leq n < w}, \]
   where $\tilde{\mu}$, $\tilde{\nu}$ and $\tilde{\nu}_k$ are given by Lemma 2.21(c), to compute a vector $(f_1, \ldots, f_\sigma)$ of approximate power series solutions of $\tilde{L}z = 0$.
4. For $i = 1, \ldots, \sigma$, call Algorithm 3 to compute $\tilde{n}_i = \max(0, N(s + n) - \lfloor \tilde{\nu} \rfloor)$ additional terms of $f_i$, thus extending it to a truncated power series solution $\hat{z}_i = z_0 + \cdots + z_N(s+n)x^{N(s+n)}$ of $\tilde{L}$.
5. Return $(\hat{y}_1, \ldots, \hat{y}_\sigma)$ where $\hat{y}_i = z_0 x^{-s} + z_1 x^{-s+1/N} + \cdots + z_N(s+n)x^n$.

\textbf{Algorithm 7}. Solving a Mahler equation in $\mathbb{K}((x^{1/N}))$.

\textbf{Proof}. Observe that $qc$ is equal to the common value on $S_0$ of $pU + qV$. Since the endpoints of $S_0$ have integer coordinates, this value is an integer, and hence the coefficients of $L$ are Laurent polynomials. The transformation $[\phi]$ of the Newton plane maps segments of slope $s$ to segments of slope $(\alpha + \beta s)/(1 + 0 \cdot s) = p + qs$, and in particular maps $S_0$ to a horizontal segment. By the choice of $c$, that segment lies on the $U$-axis. Since $q > 0$, images by $[\phi]$ of points above $S_0$ lie above $[\phi](S_0)$. As monomials of $L$ correspond to points lying on or above $S_0$, their images by $\phi$ are monomials of nonnegative degree. This proves assertion (a). It follows that $\tilde{L}$ has a Newton diagram in the sense of our definition, and it is then clear this Newton diagram is as stated by (b). The expressions of $\tilde{\nu}_k$ and $\tilde{c}_k$ in (c) are a consequence of (2.13), using again the positivity of $\beta$. Those of $\tilde{\nu}$ and $\tilde{\mu}$ follow. We already observed that $\tilde{\nu}_k \geq 0$. Finally, $-\tilde{\nu}$ and $\tilde{\mu}$ are, respectively, the slope and $V$-intercept of the leftmost edge of the lower Newton polygon $\tilde{N}$ of $\tilde{L}$. Since $\tilde{N}$ has a horizontal edge, $\tilde{\nu}$ and $\tilde{\mu}$ are nonnegative.

\textbf{Example 2.22}. Consider again the Mahler operator $L$ in (2.1) treated for $b = 3$ in §2.1. We already observed that the slopes of the Newton polygon of $L$ are $-3$ and $1/2$ and that they are admissible, and, in §2.1, we performed the transformation (2.13) for the parameters $\alpha = -1$, $\beta = 2$, and $\gamma = -3$, to obtain the operator $\tilde{L}$ in (2.5). The slopes of the Newton polygon of $\tilde{L}$ are $-7$ and $0$ and are both admissible.

\textbf{Theorem 2.23}. Algorithm 7 runs in
\[ O(r^2 M(Nd) + rN(d^2 + (r + d)n)) = \tilde{O}(r^2 Nd (d + n)) \text{ ops} \]
(assuming a softly linear-time polynomial multiplication) and computes the truncation to order $O(x^{n+1})$ of a basis of solutions of (EQN) in $\mathbb{K}((x^{1/N}))$. 


Proof. The discussion at the beginning of this section shows that \( z(x) \in \mathbb{K}((x^{1/\ast})) \) is a solution of the operator \( \hat{L} \) computed at step 2 if and only if \( y(x) = x^{-s}z(x^{1/N}) \) is a solution of \( L \). By Lemma 2.2 and the choice of \( s \), solutions of \( L \) in \( \mathbb{K}((x^{1/N})) \) have valuation at least \(-s\), and hence correspond to solutions of \( \hat{L} \) lying in \( \mathbb{K}[x] \). Since the mapping \( z \mapsto y \) is linear and invertible, a basis of solutions of \( \hat{L} \) in \( \mathbb{K}[x] \) provides a basis of solutions of \( L \) in \( \mathbb{K}((x^{1/N})) \).

Let \( S_0 \) be the edge of the Newton polygon of \( L \) considered at step 1, so that the notation of the algorithm agrees with that of Lemma 2.21. Lemma 2.21(c) then provides expressions various parameters associated to \( \hat{L} \) in terms of \( s, c, \) and quantities that can be read off \( L \). Since \( \hat{d} \) is nonnegative, Proposition 2.12 applies and shows that step 3 computes a basis \( (f_1, \ldots, f_r) \) of the space of approximate solutions of \( \hat{L} \) in \( \mathbb{K}[x] \) in \( O(rd\hat{v}_0 + r^2 M(\hat{v}_0)) \) ops. Denote by \((z_1, \ldots, z_r)\) the basis of power series solutions of \( L \) such that each \( z_i \) extends \( f_i \). Then, according to Proposition 2.13, the series \( \hat{z}_i \) computed at step 4 satisfy \( \hat{z}_i = z_i + O(x^{Nr/(r+d)n}) \), and their computation takes \( O(\sigma(r+d)n) \) ops. Finally, the truncated Puiseux series returned by the algorithm satisfy \( \hat{y}_i = x^{-s}\hat{z}_i(x^{1/N}) \), hence are truncations of elements of a basis of solutions of \( \hat{L} \) in \( \mathbb{K}((x^{1/N})) \).

Steps other than 3 and 4 do not perform any operation in \( \mathbb{K} \), so that the cost in ops of the algorithm is concentrated in those two steps. Let \((b^{k_1}, v_{k_1})\) and \((b^{k_2}, v_{k_2})\) with \( k_1 < k_2 \) be the endpoints of \( S_0 \), so that

\[
qc = pb^{k_1} + qv_{k_1} = pb^{k_2} + qv_{k_2}.
\]

Lemma 2.21(c) gives \( \hat{v}_0 = qv_{k_1} + p - qc \). If \( p \geq 0 \), then (2.15) implies \( qc \geq p \) and hence \( \hat{v}_0 \leq qv_0 \leq N \hat{d} \). If, now, \( p < 0 \), first observe that since \( b^{k_2} \geq 2b^{k_1} \), we have

\[-pb^{k_1} \leq -p(b^{k_2} - b^{k_1}) = q(v_{k_2} - v_{k_1}).\]

It follows that \( -qc = -pb^{k_1} - qv_{k_1} \leq qv_{k_2} \), whence \( \hat{v}_0 \leq q(v_{k_0} + v_{k_2}) \leq 2N \hat{d} \). In both cases, we have proved that \( \hat{v}_0 = O(\hat{N} \hat{d}) \).

The complexity estimate for step 3 thus rewrites as \( O(r \hat{N} \hat{d}^2 + r^2 M(\hat{N} \hat{d})) \) ops. As \( s \leq d \) (because all slopes of the Newton polygon are bounded by \( d \) in absolute value) and \( \sigma \leq r \), that of step 4 becomes \( O(r \hat{N}(r+d)(d+n)) \) ops. The total running time is therefore \( O(r^2 M(\hat{N} \hat{d}) + r \hat{N}(d^2 + (r + d)n)) \) ops.

Recall that \( Q \) denotes the set of denominators \( q \) of slopes, written in lowest terms, of admissible edges of \( \hat{N} \) such that \( q \wedge b = 1 \).

**Corollary 2.24.** Algorithm 7 with \( N \) set to the lcm of elements in \( Q \), returns the truncation to order \( O(x^{n+1}) \) of a basis of solutions of \((EQN)\) in \( \mathbb{K}((x^{1/\ast})) \) in \( \hat{O}(r^2b^d(d+n)) \) ops, assuming \( M(k) = \hat{O}(k) \).

**Proof.** This follows by combining Proposition 2.19 with Theorem 2.23. 

**Example 2.25.** With \( b = 3 \), let us consider the order \( r = 11 \) Mahler operator

\[
L = x^{568} - (x^{1218} + x^{1705})M + x^{3655}M^2 - (x^{162} - x^{10962})M^3 \\
+ (1 + x^{487} - x^{4104} - x^{4536} - x^{32887})M^4 - (x - x^{11826} - x^{12313} - x^{13122} - x^{13609})M^5 \\
- (1 + x^{35479} + x^{39367})M^6 + (x + x^{95634} - x^{106434} - x^{118098})M^7 \\
- (x^{286416} + x^{286903} - x^{319303} - x^{354295})M^8 + x^{859249}M^9 \\
+ x^{257774}M^{10} - x^{773323}M^{11}.
\]

Its associated parameters are \( w = 0 \), \( \hat{v}_0 = 568 \), and a Newton polygon made from five segments, all admissible, with slopes \(-203/13, -3, 0, 1/1458, \) and \( 221/5 \). Except for
We want to find a basis of Puiseux solutions for \(1458 = 2 \cdot 3^6\), the denominators are coprime with \(b = 3\) and their \(\text{lcm}\) is \(N = 65\). The rightmost slope is \(s = 221/5\) and we perform the change of variables of Algorithm 7 with \(\alpha = -2873\), \(\beta = 65\), hence \(\gamma = -6283186\) and this provides us with the new operator

\[
\tilde{L} = t^{6317233} - (t^{6353737} + t^{6385392})M + t^{6494904}M^2 - (t^{6216145} - t^{6918145})M^3 \\
+ (t^{6050473} + t^{6082128} - t^{6317233} - t^{6345313} - t^{8188128})M^4 \\
- (t^{5585112} - t^{6353737} - t^{6385392} - t^{6437977} - t^{6469632})M^5 \\
- (t^{4188769} - t^{6494904} - t^{7476241})M^6 + (1 + t^{6216145} - t^{6918145} - t^{7676305})M^7 \\
- (t^{6050473} + t^{6082128} - t^{8188128} - t^{10462008})M^8 + t^{5585112}M^9 + t^{4188769}M^{10} - M^{11}.
\]

We want to find a basis of Puiseux solutions for \(\tilde{L}\) with a precision \(O(x^n)\) where \(n = 10^6\). According to Algorithm 7, this leads us to compute a basis of formal series solutions for \(\tilde{L}\) with a precision \(O(x^n)\) where \(\tilde{n} = 65002873\). We first apply Algorithm 4 with \(\tilde{\nu} = 3888\), \(\tilde{\mu} = 6321121\). The computation shows that the space of solutions has dimension 2. We extend the solutions to the requested precision by Algorithm 3 and we obtain a basis of formal series solutions

\[
\tilde{f}_1(t) = 1 + t^{28080} + t^{657072} + t^{2274480} + t^{2305260} + t^{17639856} + t^{53222832} \\
+ t^{53250912} + t^{62068032} + O(t^{65002873})
\]

\[
\tilde{f}_2(t) = t^{3888} + t^{314928} + t^{343008} + t^{6821028} + t^{25509168} + t^{25537248} \\
+ t^{27783648} + t^{27811728} + O(t^{65002873}).
\]

Reversing the change of variable, we find the basis

\[
f_1(x) = x - \frac{221}{5} + x^\frac{1039}{5} + x^\frac{50233}{5} + x^\frac{174739}{5} + x^\frac{17639856}{5} + x^\frac{1356691}{5} + x^\frac{4093843}{5} \\
+ x^\frac{4096803}{5} + x^\frac{27783648}{5} + O(x^{1000000})
\]

\[
f_2(x) = x^\frac{203}{13} + x^\frac{9241}{13} + x^\frac{6827}{13} + x^\frac{1831451}{13} + x^\frac{5101259}{13} + x^\frac{5106875}{13} \\
+ x^\frac{5256155}{13} + x^\frac{5561771}{13} + O(x^{1000000}).
\]

These truncated series satisfy \(Lf_1 = O(x^e)\), \(Lf_2 = O(x^e)\) with \(e = v_0 + n = 1000568\).

3. Rational solutions

We now turn to the computation of rational function solutions of Mahler equations of the form (EQN). Our algorithm follows a classical pattern: it first computes a denominator bound, that is, a polynomial that the denominator of any (irreducible) rational solution must divide. Then it makes a change of unknown functions and computes the possible numerators using the algorithm of §2.6. As is usual with other functional equations, the denominator bound is obtained by analyzing the action of the operator \(L\) on zeros and poles of the functions it is applied to.
3.1. **Denominator bounds: setting.** We will call a rational function \( p/(x^\nu q) \) in lowest terms if it satisfies the following conditions: \( \nu \geq 0; \ p, q \in \mathbb{K}[x] \) are coprime polynomials; \( q(0) \neq 0 \); and \( p(0) \) can be zero only if \( \nu = 0 \).

Consider a rational solution \( p/(x^\nu q) \) of (EQN), written in lowest terms. We already know from Lemma 2.2 that \( \nu \leq v_r/(b^r - b^{r-1}) \), so we are left with the problem of finding a multiple of \( q \).

Write \( Ta = \sqrt[r-1]{M^i}a \). We will freely use the fact that \( T(ab) \mid (Ta)(Tb) \) for all \( a \) and \( b \). For any \( j \) between 0 and \( r \), multiplying the equation

\[
\ell_r(x)M^\nu y + \cdots + \ell_1(x)My + \ell_0(x)y = 0,
\]

by \((M^\nu x^\nu)(M^\nu q)\) \( \bigvee_{i\neq j} M^\nu q \) and reducing modulo \( M^\nu q \) yields

\[
M^\nu q \mid x^{(b^r - b^0)} \ell_j(M^\nu p) \bigvee_{i\neq j} M^\nu q.
\]

As \( q \) is coprime with \( p \) and \( q(0) \neq 0 \), Equation (3.1) with \( j = r \) implies

\[
M^\nu q \mid \ell_r Tq.
\]

This relation is our starting point for computing a polynomial \( q^* \), depending only on \( \ell_r \), such that \( q \mid q^* \).

The algorithm for this task, presented in §3.3, operates with polynomials over \( \mathbb{K} \), but it may be helpful in order to get an intuition to first consider the case \( \mathbb{K} = \mathbb{C} \). Assume for simplicity that \( q \) is squarefree. Equation (3.2) then says that, if \( \alpha \) is a zero of \( q \), each of its \( b^r \)th roots is either a \( b^k \)th root with \( k < r \) of some zero of \( q \) or a zero of \( \ell_r \). Thus, when \( \alpha \) is not a root of unity, its \( b^r \)th roots are either \( \ell_r \) or roots of lower order of some other zero of \( q \), whose \( b^r \)th roots then satisfy the same property. (Compare Lemma 3.4 below.) As \( q \) has finitely many zeros, this cannot continue indefinitely, so, in this case, we will eventually find a zero \( \alpha \) whose \( b^r \)th roots are zeros of \( \ell_r \). A difficulty arises when \( \alpha \) is a root of unity, but then at most one of its \( b^\nu \)th roots can be part of a cycle of the map \( \zeta \mapsto \zeta^b \) (cf. Lemma 3.6), and a closer examination shows that the \( b - 1 \) other roots behave essentially like non-roots of unity.

3.2. **Properties of the Mahler and Graeffe operators.** Going back to the general case, and before making the reasoning sketched above more precise, let us state a few properties of the action of \( M \) on polynomials. Besides \( M \), we consider the Graeffe operator defined by

\[
G : \mathbb{K}[x] \to \mathbb{K}[x], \quad p \mapsto \text{Res}_y(y^b - x, p(y)).
\]

In other words, \( Gp \) is the product \( p(x^{1/b})p(\zeta x^{1/b}) \cdots p(\zeta^{b-1} x^{1/b}) \) for any primitive \( b^\nu \)th root of unity \( \zeta \). While \( M \) maps a polynomial \( p \) to a polynomial whose complex zeros are the \( b^\nu \)th roots of the zeros of \( p \), the zeros of \( Gp \) are the \( b^\nu \)th powers of the zeros of \( p \).

As a direct consequence of the definitions, \( M \) and \( G \) act on degrees by:

\[
\deg M p = b \deg p, \quad \deg G p = \deg p.
\]

Some other elementary properties that will be useful in the sequel are as follows.

**Lemma 3.1.** For any nonzero \( i \in \mathbb{N} \), the following relations between \( M \) and \( G \) hold for all \( p, q \in \mathbb{K}[x]:

(a) \( G^i M^j p = p^{b^i} \),
(b) $p | M^i G^i p$,
(c) $p | q \iff M^i p | M^i q$.

Proof. The case $i > 1$ reduces to the case $i = 1$ by changing the radix, since $M^i$ (resp. $G^i$) is nothing but the Mahler (resp. Graeffe) operator of radix $b^i$, so we set $i = 1$. The assertions (a) and (b) are direct consequences of the definition of $G$ as a resultant. The direct implication in (c) is clear. For the converse, write the Euclidean division $q = up + v$. If $Mq = sMp$ for some $s \in \mathbb{K}[x]$, then $(Mu)(Mp) + (Mv) = sMp$, whence $Mv = 0$ since $\deg Mv < \deg Mp$. \hfill \Box

Lemma 3.2. If $p \in \mathbb{K}[x]$ is monic, irreducible and $i \in \mathbb{N}$, then $G^i p = q^e$ for some monic, irreducible $q \in \mathbb{K}[x]$ and $e \in \mathbb{N}$. Furthermore, $G^i p = p$ if and only if $p$ divides $M^i p$.

Proof. To prove the first point, consider the factorization $G^i p = c q_1^{e_1} \cdots q_s^{e_s}$ of $G^i p$ for monic, irreducible, pairwise coprime $q_j$ and a nonzero $c \in \mathbb{K}$. Because of Lemma 3.1(c), the polynomials $M^i q_1^{e_1}, \ldots, M^i q_s^{e_s}$ are pairwise coprime. We have

$$M^i G^i p = c (M^i q_1^{e_1}) \cdots (M^i q_s^{e_s}),$$

and, by Lemma 3.1(b), $p | M^i q_j^{e_j}$ for some $j$. It follows that $G^i p | G^i M^i q_j^{e_j} = q_j^{e_j b^i}$ by Lemma 3.1(a), proving the first point.

Now if $p | M^i p$, then $G^i p | p^{b^i}$, and necessarily there is $e \in \mathbb{N}$ such that $G^i p = p^e$. In fact, $e = 1$ and $G^i p = p$ as $G^i p$ and $p$ have the same degree and $p$ is irreducible. Conversely, if $G^i p = p$, then $p$ divides $M^i p$ by Lemma 3.1(b). \hfill \Box

Lemma 3.3. Let $f \in \mathbb{K}[x]$ be a nonconstant polynomial with $f(0) \neq 0$. If $f$ and its derivative $f'$ are coprime, so are $M^i f$ and $(M^i f)'$.

Proof. Assume $f \land f' = 1$. Applying $M$ to a Bézout relation shows that $Mf \land M(f') = 1$. Now, $(Mf)' = bx^{k-1}M(f')$, so a common factor $s$ of $Mf$ and $(Mf)'$ must divide $x$. As $x$ cannot divide $Mf$ because $x \nmid f$, the only possibility is that $s$ be a constant. \hfill \Box

The following lemma generalizes the fact that the iterated $b$th roots of a complex number $\alpha \neq 0$ are all distinct, except in some cases where $\alpha$ is a root of unity.

Lemma 3.4. Let $p \in \mathbb{K}[x]$ be monic and irreducible. For general $\mathbb{K}$, $M^i p$ and $M^j p$ are coprime for all $i > j \geq 0$ if none of the $G^i p$ for $i \geq 1$ is equal to $p$. When $\mathbb{K} = \mathbb{Q}$, the same conclusion holds if $Gp$ is not equal to $p$.

Proof. We proceed by contraposition, assuming the negation of the common conclusion: for monic, irreducible $p$, assume $M^i p \land M^j p \neq 1$ for some $i > j \geq 0$. Set $k = i - j \geq 1$. Lemma 3.1(c) implies that $M^k p$ and $p$ are not coprime. Then $p$ divides $M^k p$ and Lemma 3.2 implies that $G^k p = p$. This proves the result for general $\mathbb{K}$. For $\mathbb{K} = \mathbb{Q}$, a further consequence is that the map $\alpha \mapsto \alpha^{b^k}$ is a permutation of the roots of $p$ in $\mathbb{Q}$. Hence, all roots of $p$ satisfy $\alpha^{b^e} = \alpha$ for some power $B = b^e$ of $b$, with $e > 0$. This means that $p$ divides $x^{b^e} - x$. If $p = x$, $Gp = p$; otherwise, $p$ is a cyclotomic polynomial $\Phi_a$ with $a | b^e - 1$, so $a \land b = 1$. Applying the formula in [8, Prop. 4 p. 14] yields $M\Phi_a = \prod_{b \nmid a} \Phi_{ab}$, so that $p$ divides $Mp$. Lemma 3.2 now implies $Gp = p$ again, completing the proof. \hfill \Box
Remark 3.5. Over a general subfield \( \mathbb{K} \subset \mathbb{C} \), the cyclotomic polynomial \( \Phi_a \) factors as \( \Phi_a = \Psi_1 \cdots \Psi_s \) and \( G \) acts as a cyclic permutation of the \( \Psi_i \). See also [8, Chap. 1] for a detailed description of the case \( a \wedge b \neq 1 \).

Lemma 3.4 does not apply to polynomials \( p \) that are part of a cycle of the map \( G \). We can always reduce to its hypothesis \( G^i p \neq p \) for nonzero \( i \), by “stepping back one step” in the graph of \( G \).

Lemma 3.6. Let \( f \in \mathbb{K}[x] \) be a nonconstant polynomial with \( f(0) \neq 0 \). There exists a monic irreducible factor \( q \in \mathbb{K}[x] \) of \( Mf \) such that \( G^k q \neq q \) for all nonzero \( k \in \mathbb{N} \).

Proof. Choose a monic irreducible factor \( p \) of \( f \) and write \( Mp = q_1 \cdots q_s \) for monic irreducible \( q_i \). By contradiction, assume that for each \( i \), there is some nonzero \( k_i \) for which \( G^{k_i} q_i = q_i \). It follows that for \( k = k_1 \cdots k_s \) and all \( i \), \( G^k q_i = q_i \). Lemma 3.1(a) implies \( p^b = (Gq_1) \cdots (Gq_s) \), and because of Lemma 3.2, for all \( i \), \( Gq_i \) is irreducible. Hence, there exist nonzero \( e_i \in \mathbb{N} \) such that \( Gq_i = p^{e_i} \), with \( b = e_1 + \cdots + e_s \). Therefore, for each \( i \), \( q_i = G^{k_i - 1} p^{e_i} \), so that, as \( q_i \) is irreducible, \( e_i = 1 \), and thus all \( q_i \) are equal to some same monic irreducible \( \tilde{q} \). It follows that \( Mp = \tilde{q} \). As \( p \) is irreducible, Lemma 3.3 applies to show that \( Mp \wedge (Mp)' = 1 \), which is impossible. The result follows by setting \( q = q_i \) for a suitable \( i \).

Example 3.7. To suggest the graph structures induced by the Mahler and Graeffe operators, we depict on Figure 4 the graph of the map that maps a monic, irreducible \( p \) to the unique monic, irreducible \( q \) such that \( Gp \) is some power of \( q \). Applying \( M \) to some vertex \( p \) in the graph results in the product of all antecedents under the map. For example, \( M(x - 2^6) = (x - 2)(x + 2)(x^2 - 2x + 4)(x^2 + 2x + 4) \),
We prove by contradiction that \( \deg M = \Phi_a + 2\Phi_2 + \Phi_3 + \Phi_6 \). In the second example, \( \Phi_a \) appears to the right as a consequence of it being mapped to itself by \( G \).

The depicted case, \( b = 6 \), is typical for \( \mathbb{Q}[x] \). In particular, all cycles have length 1 as a consequence of the second part of Lemma 3.4.

### 3.3. Denominator bounds: algorithm

Armed with the previous lemmas, we can now prove the key result that leads to our main denominator bound.

**Proposition 3.8.** Let \( q \in \mathbb{K}[x] \) be a nonconstant polynomial such that \( x \nmid q \) and \( M^r q \mid \ell Tq \). Then there exists a nonconstant \( u \in \mathbb{K}[x] \) such that:

- either \( M^r u \mid \ell \),
- or \( M^{r-1} u \mid \ell \) and \( q \mid Gu \).

**Proof.** Consider a monic irreducible \( f \) such that \( f^m \mid q \) for some \( m > 0 \). Two cases are possible:

1. If \( G^i f \neq f \) for all \( i > 0 \), set \( \tilde{f} \) to \( f \).
2. Otherwise, Lemma 3.6 provides a monic irreducible factor \( \tilde{f} \in \mathbb{K}[x] \) of \( Mf \) such that \( G^i \tilde{f} \neq \tilde{f} \) for \( i > 0 \).

In both cases, \( G^i \tilde{f} \neq \tilde{f} \) for all \( i > 0 \), so by Lemma 3.4, the \( M^i \tilde{f} \) for \( i \in \mathbb{N} \) are pairwise coprime. Since \( q \) has finitely many factors, \( M^i \tilde{f} \land q \neq 1 \) for at most finitely many \( i \). Again, two situations are possible:

1. There exists a maximum \( j \geq 0 \) with \( M^j \tilde{f} \land q \neq 1 \). Set \( g \) to a monic irreducible factor \( \tilde{f} \in \mathbb{K}[x] \) of \( Mf \) and, by maximality of \( j \), \( M^i g \) is coprime with \( q \) for all \( i \) satisfying \( 0 \leq i < r \).
2. The polynomials \( M^i \tilde{f} \) are coprime with \( q \) for all \( i \in \mathbb{N} \). This is only possible in case (I), since in case (II), \( \tilde{f} \land q \neq 1 \). Set \( g \) to \( \tilde{f} \) and let \( s = r - 1 \). Then, \( M^i g = M^{s+i} \tilde{f} \land M^r \tilde{f} \land M^r q \), and in this case as well \( M^{s+i} g = M^{s+i-1} \tilde{f} \) is coprime with \( q \) for all \( i \) satisfying \( 0 \leq i < r \).

In both cases, \( M^s g \mid M^r q \) and \( M^{s+i} g \land q = 1 \) whenever \( 0 \leq i < r \).

Let \( h^k \mid M^r g \) for a monic irreducible \( h \in \mathbb{K}[x] \) and \( k > 0 \), so that \( h^k \mid M^r q \mid \ell Tq \). We prove by contradiction that \( h \) is coprime with \( Tq \): suppose there exists some \( i \) satisfying \( 0 \leq i < r \) such that \( h \mid M^i q \). Then, \( G^i h \) divides both \( G^i M^i q \) and \( G^i M^r g \), which, upon applying Lemma 3.1(a), are equal to powers of \( q \) and \( M^{s+i} g \), respectively. This contradicts the coprimality of \( q \) and \( M^{s+i} g \). We conclude that \( h^k \mid \ell \), and, by considering all \( h^k \mid M^s g \), it follows that \( M^s g \mid \ell \).

Additionally, in case (β), \( Gg = G \tilde{f} \mid GMf = f^b \), so that \( Gg \) is a power of \( f \), hence \( f \mid Gg \), and next \( f^m \mid Gg^m \).

Now let \( q = \prod_k f_k^{m_k} \) be the irreducible factorization of \( q \), and denote by \( \tilde{f}_k \), resp. \( g_k \), the polynomials \( f_k \), resp. \( g_k \), constructed above starting from \( f_k \). If any of the \( \tilde{f}_k \) falls into case (α), the corresponding \( g_k \) satisfies \( M^r g_k \mid \ell \). If all \( \tilde{f}_k \) fall into case (β), the \( \tilde{f}_k \) are pairwise coprime, and so are the \( g_k \). It follows that all \( M^{r-1} g_k \) divide \( \ell \) and are pairwise coprime, so that, finally, the product \( u = \prod_k g_k^{m_k} \) satisfies \( M^{r-1} u \mid \ell \) and \( q \mid Gu \). \( \square \)

**Corollary 3.9.** When \( d < b^{-1} \), Eq. (EQN) has no nonconstant rational solution.

**Proof.** With the notation above, Lemma 2.2 implies \( \check{v} = 0 \). If a nonconstant \( q \) could satisfy Eq. (3.2), Proposition 3.8 would apply, inducing the contradiction \( b^{-1} \leq \deg \ell_r \leq d \). So \( q \) is constant, and Lemma 2.5 applies and proves \( p \) is constant. \( \square \)
Input: A linear Mahler equation of the form \((\text{eqn})\).

Output: A polynomial \(q^* \in \mathbb{K}[x]\).

(1) Set \(\ell := \ell_r\), then repeat for \(k = 1, 2, \ldots\):
   (a) write \(\ell = \sum_{i=0}^{b^r-1} x^i M^r f_i\) with \(f_i \in \mathbb{K}[x]\);
   (b) set \(u_k := \bigwedge_{i=0}^{b^r-1} f_i\);
   (c) set \(\ell := (\ell/M^r u_k) \bigwedge_{i=0}^{r-1} M^i u_k\)
       until \(\text{deg } u_k = 0\), at which point set \(t = k - 1\).

(2) Set \(\tilde{u} := \bigwedge_{i=0}^{b^{r-1}-1} f_i\) where \(\ell = \sum_{i=0}^{b^{r-1}-1} x^i M^{r-1} f_i\).

(3) Return \(u_1 \cdots u_t (G\tilde{u})\).

Algorithm 8. Obtain a denominator bound from \(\ell_r\).

Proposition 3.8 forms the basis of Algorithm 8, which repeatedly searches for factors of the form \(M^r u\) to “be removed” from \(\ell_r\) (while “adding back” other factors of strictly smaller degree) and accumulates the corresponding \(u\) into the denominator bound. The update of \(\ell\) at step 1c of each loop iteration can be viewed as a change of unknown functions of the form \(y = \tilde{y}/u_k\) in \((\text{eqn})\). The search for factors of the form \(M^r u\), respectively \(M^{r-1} \tilde{u}\), uses the following property (for radix \(b^r\), resp. \(b^{r-1}\)).

Lemma 3.10. Let \(f_0, \ldots, f_{b-1}, u \in \mathbb{K}[x]\). The polynomial \(\ell = M f_0 + x M f_1 + \cdots + x^{b-1} M f_{b-1}\) is divisible by \(M u\) if and only if \(f_0, \ldots, f_{b-1}\) are all divisible by \(u\).

Proof. The “if” part is clear. Conversely, fix \(i < b\), and assume that \(M u \mid \ell\). Let \(\omega\) be a primitive \(b\)th root of unity. Then, \(M u = (M u)(\omega^j x) \mid \ell(\omega^j x)\) for all \(j\), hence \(M u\) divides \(\sum_{j=0}^{b-1} \omega^{-ij} \ell(\omega^j x) = b x^i M f_i\).

As \(M u \in \mathbb{K}[x^b]\) and \(i < b\), this implies \(M u \mid M f_i\), and \(u \mid f_i\) by Lemma 3.1(c). \(\square\)

Example 3.11. In this example, we let \(b = 3\) and use Algorithm 8 to analyze the potential poles in rational-function solutions of an operator

\[ L = (p_1(x) \cdots p_6(x)) M^2 + \cdots, \]

where the \(p_i\) are polynomials to be found in Figure 5 and the coefficients of \(M^1\) and \(M^0\) will be disclosed below. In the figure and this example, polynomials of large size are truncated to their first few monomials, and in most cases, we write them in factored form, although polynomials are manipulated in expanded form in the actual algorithm.

Following Algorithm 8, we set \(\ell = p_1 \cdots p_6\). Step 1 is motivated by the first case in Proposition 3.8: it strives to solve (3.2) by finding a factor \(u\) of \(q\) such that \(M^2 u \mid \ell\). For each \(i\), the only monic, irreducible candidate factor of \(u\) that can “cover” \(p_i\) upon application of \(M^2\) is the polynomial \(p_{i,2}\) in the figure. However, \(M^2 p_{i,2}\) consists of all factors on the level of \(p_i\) with same ancestor \(p_{i,2}\). So, for example, \(M_2 p_{1,2} = p_1\) and \(p_{1,2}\) can be part of \(u\), whereas \(M_2 p_{6,2}\) is a strict multiple of \(p_6\) so that \(p_{6,2}\) cannot be made part of \(u\). As a matter of fact, for \(k = 1\) in the
loop, the algorithm finds \( u_1 = p_{1,2} p_{2,2} p_{4,2} p_{5,2} \) at step 1b, after rewriting \( \ell \) in the form

\[
\ell = -M^2 \left( (2x - 1)(x^2 - x + 1)(x^2 - x - 1)(x^6 - x^3 - 1)(9x^5 - 133x^4 + \cdots) \right) + \cdots
\]

\[
x^8 M^2 \left( 2(2x - 1)(x^2 - x + 1)(x^2 - x - 1)(x^6 - x^3 - 1)(5x^4 - 74x^3 + \cdots) \right)
\]

at step 1a. Step 1c resets \( \ell \) to a polynomial that factors into

\[ p_{1,1} p_{1,2} p_{2,1} p_{2,2} p_{3} p_{4} p_{5,2} p_{4,2} p_{6} \]

Following the same approach for \( k = 2 \), a new phenomenon occurs because of the loops in the graph: the candidate factor \( p_{2,3} \) that would “cover” \( p_{2,1} \) appears in its own tree on the same level as \( p_{2,1} \), and thus has to be rejected. It follows that the algorithm finds \( u_2 = p_{3,2} p_{4,2} \) at step 1b, after rewriting \( \ell \) in the form

\[
\ell = -M^2 \left( (x^2 - 4x - 1)(x^2 - x - 1)(248x^5 - 5615x^4 + \cdots) \right) + \cdots +
\]

\[
x^8 M^2 \left( (x^2 - 4x - 1)(x^2 - x - 1)(532x^4 - 6211x^3 + \cdots) \right)
\]

at step 1a. Step 1c resets \( \ell \) to a polynomial that factors into

\[ p_{1,1} p_{1,2} p_{2,1} p_{2,2} p_{3,1} p_{5,2} p_{4,2} p_{3,2} p_{6} \]
Following the same approach for \( k = 3 \) leads to \( u_3 = 1 \): no further factor \( u \) of \( q \) exists and helps solving Eq. (3.2) by ensuring \( M^2u \mid \ell \).

This leads to step 2, which is motivated by the second case in Proposition 3.8: Eq. (3.2) now implies \( M^2q \mid q \wedge Mq \), which is solved by finding \( \tilde{u} \) such that \( M\tilde{u} \mid \ell \). A difference to step 1 is that at step 2, candidates are looked for just \( 2 - 1 = 1 \) level above the factors to be “covered”. A similar calculation as previously explains that the algorithm finds \( \tilde{u} = p_{1,2}p_{2,2}p_{3,2}p_{4,2} \), after rewriting \( \ell \) in the form

\[
\ell = M^2((2x - 1)(x^2 - x + 1)(x^2 - 4x - 1)(x^2 - x - 1)(181x^{13} - 1198x^{12} + \cdots)) \\
- xM^2((2x - 1)(x^2 - x + 1)(x^2 - 4x - 1)(x^2 - x - 1)(44x^{13} - 623x^{12} + \cdots)) \\
+ x^2M^2((2x - 1)(x^2 - x + 1)(x^2 - 4x - 1)(x^2 - x - 1)(4x^{13} - 382x^{12} + \cdots)).
\]

From these factors, only \( p_{2,2} \) is cyclotomic. But as the algorithm does not factor polynomials, the other factors cannot be discarded.

At step 3, the algorithm returns the bound

\[
q^* = u_1u_2G\tilde{u} = p_{1,2}p_{2,2}p_{3,2}p_{4,2}
\]

where the “+1” indicate factors that could have been saved if a cyclotomic test had been available. The operator \( L \) was indeed constructed so as to admit the two explicit rational solutions

\[
\frac{2x}{(2x - 1)(x^2 - x - 1)} \quad \text{and} \quad \frac{x - 3}{(x^2 - x + 1)(x^2 - 4x - 1)(x^6 - x^3 - 1)},
\]

whose denominators are effectively “covered” by \( q^* \).

We remark that, during the steps of the algorithm, the degree of \( \ell \) has dropped from its initial value 145 down to 84, then to 62.

**Example 3.12.** Let \( b = 3 \) and let us consider the Mahler equation

\[
L = (2x^4 - x^3 - x + 3)(2x^9 - 1)(x^{18} - x^9 - 1) M^2 \\
- (x^2 + 1)(2x^3 - 1)(x^4 + 1)(x^6 - x^3 - 1)(2x^{10} - x^9 - x + 3) M \\
+ x^2(2x - 1)(x^2 + x + 1)(x^2 - x + 1)(x^2 - x - 1)(2x^{12} - x^9 - x^3 + 3).
\]

Following Algorithm 8, we expand \((2x^4 - x^3 - x + 3)(2x^9 - 1)(x^{18} - x^9 - 1)\) to get \( \ell \), which step 1a rewrites

\[
\ell = M^2(6x^3 - 9x^2 - 3x + 3) + xM^2(-2x^3 + 3x^2 + x - 1) + \\
x^3M^2(-2x^3 + 3x^2 + x - 1) + x^4M^2(4x^3 - 6x^2 - 2x + 2).
\]

(That is, \( f_2 = f_5 = f_6 = f_7 = f_8 = 0 \).) We get \( u_1 = 2x^3 - 3x^2 - x + 1 \), which factors into \((2x - 1)(x^2 - x - 1)\). Step 1c resets \( \ell \) to a polynomial that factors into \((2x - 1)(x^2 - x - 1)(2x^3 - 1)(2x^4 - x^3 - x + 3)(x^6 - x^3 - 1)\). Expanding \( \ell \) as in step 1a, we now find

\[
\ell = M^2(3 - 10x) + xM^2(-4 - 15x) + x^2M^2(-8 - 19x) + \\
x^3M^2(5 + 40x) + x^4M^2(5 - 10x) + x^5M^2(2x + 9) + \\
x^6M^2(-25 - 16x) + x^7M^2(15 + 8x) + x^8M^2(23).
\]
so that \( u_2 = 1 \). We pass to step 2, which expands \( \ell \) in the form

\[
\ell = M(-16x^5 + 40x^4 - 10x^3 - 25x^2 + 5x + 3) + xM(8x^5 - 10x^4 - 15x^3 + 15x^2 + 5x - 4) + x^2M(2x^4 - 19x^3 + 23x^2 + 9x - 8),
\]

and \( \tilde{u} = (2x-1)(x^2-x-1) \). So, \( q^* = u_1 G\tilde{u} = (2x-1)(x^2-x-1)(8x-1)(x^2-4x-1) \). This means that if \( y = p/(x^q) \) is solution of \( L_y = 0 \), where \( v \geq 0 \) and \( p, q \in \mathbb{K}[x] \) satisfy \( x \cap q = p \cap q = p \cap x^v = 1 \), then \( q \) divides \( q^* \). Using the results of §2.2, we find that 0 could not be a pole of a solution in \( \mathbb{K}(x) \) and therefore \( \tilde{v} = 0 \). Consequently, \( q^* \) is a denominator bound.

**Proposition 3.13.** Algorithm 8 runs in \( O((\deg \ell_r) M(d) \log d) \) ops if \( b = 2 \), resp. in \( O(b^{-r} (\deg \ell_r) M(d) \log d) \) ops if \( b \geq 3 \), and computes a polynomial \( q^* \) of degree at most \( \deg \ell_r \) if \( b = 2 \), resp. at most \( (\deg \ell_r)/b^{-r} \) if \( b \geq 3 \), such that any rational function solution \( y \) of (EQN) can be written in the form \( y = p/(x^q q^*) \) for some \( p \in \mathbb{K}[x] \) and \( \tilde{v} \in \mathbb{N} \).

**Proof.** For each \( k \geq 1 \) reached by the loop 1, let \( \ell_k \) denote the value of \( \ell \) considered at step 1a, so that the value assigned at step 1c is \( \ell_{k+1} \). (In particular, \( \ell_1 = \ell_r \).

First, observe that, after step 1b in each loop iteration, \( u_k \) is by Lemma 3.10 a polynomial of maximal degree such that \( M^r u_k \mid \ell_k \). In particular, the next value, \( \ell_{k+1} \), computed at step 1c, is a polynomial. Set \( \rho = b^r - b^{-1} \), which is at least 1. Step 1c decreases the degree of \( \ell \) by

\[
\deg \ell_k - \deg \ell_{k+1} \geq \deg M^r u_k - \deg Tu_k \geq \rho \deg u_k \geq \rho.
\]

In particular, the loop terminates after at most \( \rho^{-1}(1 + \deg \ell_r) \) iterations, and therefore the whole algorithm terminates as well. Second, after step 2, \( \tilde{u} \) is similarly a polynomial of maximal degree such that \( M^{r-1} \tilde{u} \mid \ell_{t+1} \). Therefore, \( b^{-r-1} \deg \tilde{u} \) is bounded above by the degree of \( \ell_{t+1} \), so that

\[
\left( \sum_{k=1}^{t} \rho \deg u_k \right) + b^{-1} \deg \tilde{u} \leq \left( \sum_{k=1}^{t} \deg \ell_k - \deg \ell_{k+1} \right) + \deg \ell_{t+1} \leq \deg \ell_r,
\]

where \( t \) denotes, as in Algorithm 8, the last value of \( k \) for which \( \deg u_k > 0 \). The output from the algorithm is \( q^* = u_1 \cdots u_t (G\tilde{u}) \). If \( b = 2 \), then \( \rho = 1 \) and \( \deg q^* = (\sum_k \deg u_k) + \deg \tilde{u} \) is bounded by \( \deg \ell_r \); if \( b \geq 3 \), then

\[
\rho = b^{-1} \left( b - 2 + \frac{b - 2}{b - 1} \right) + \frac{1}{b - 1} \geq b^{-r-1}
\]

and \( \deg q^* \) is bounded by \( b^{-(r-1)} \deg \ell_r \).

Assume that \( p/(x^q) \) is a solution written in lowest terms. Set \( \tilde{q}_0 = q \) and, for \( k \) between 1 and \( t \), define the polynomials \( \tilde{q}_k = \tilde{q}_{k-1}/(u_k \land \tilde{q}_{k-1}) \). Let us prove by an induction on \( k \) that, for \( 1 \leq k \leq t + 1 \), there exists \( \tilde{q}_{k-1} \in \mathbb{K}[x] \) such that: (i) \( x \nmid \tilde{q}_{k-1} \); (ii) \( M^r \tilde{q}_{k-1} \mid \ell_k T\tilde{q}_{k-1} \); and (iii) \( q \mid u_1 \cdots u_{k-1} \tilde{q}_{k-1} \). Initially when \( k = 1 \), we have \( \tilde{q}_0 = q \) and \( \ell_1 = \ell_r \), so the three properties hold by our assumption on a solution and Equation (3.2). Assume now that \( x \nmid \tilde{q}_{k-1} \), \( M^r \tilde{q}_{k-1} \mid \ell_k T\tilde{q}_{k-1} \) and \( q \mid u_1 \cdots u_{k-1} \tilde{q}_{k-1} \). It follows from \( \tilde{q}_{k-1} = (u_k \land \tilde{q}_{k-1}) \tilde{q}_k \mid u_k \tilde{q}_k \) that \( x \nmid \tilde{q}_k \) and \( \tilde{q}_{k-1} \mid Tu_k (T\tilde{q}_k) \). Furthermore,

\[
(M^r(u_k \land \tilde{q}_{k-1})) (M^r\tilde{q}_k) = M^r \tilde{q}_{k-1} \mid \ell_k T\tilde{q}_{k-1} \mid \ell_k (Tu_k) (T\tilde{q}_k).
\]
Write \( M^r u_k = a_k M^r (\tilde{q}_{k-1} \land u_k) \) and \( \ell_k = b_k M^r u_k \), for suitable polynomials \( a_k \) and \( b_k \). Upon division by \( M^r (u_k \land \tilde{q}_{k-1}) \), Equation (3.3) becomes
\[
M^r \tilde{q}_k | a_k b_k (Tu_k) (T \tilde{q}_k).
\]
By construction, \( a_k \) and \( M^r \tilde{q}_k \) are coprime, as they are the cofactors of \( M^r (u_k \land \tilde{q}_{k-1}) \) in, respectively, \( M^r u_k \) and \( \tilde{q}_k \), so Equation (3.4) finally becomes
\[
M^r \tilde{q}_k | \frac{\ell_k}{M^r u_k} (Tu_k) (T \tilde{q}_k) = \ell_{k+1} T \tilde{q}_k.
\]
By the divisibility assumption on \( q \) and the definition of \( \tilde{q}_k \),
\[
q | u_1 \cdots u_{k-1} \tilde{q}_{k-1} = u_1 \cdots u_{k-1} (u_k \land \tilde{q}_{k-1}) \tilde{q}_k | u_1 \cdots u_{k} \tilde{q}_k,
\]
completing the proof by induction.

The loop terminates when \( t \) no longer has any nonconstant factor of the form \( M^r u \), with \( t = t_{t+1} \). At this point, \( M^r \tilde{q}_k | \ell_{t+1} T \tilde{q}_k \) and \( q | u_1 \cdots u_{t} \tilde{q}_k \). If \( \tilde{q}_k \) is constant, then \( q | u_1 \cdots u_{t} \tilde{q}_k \). On the other hand, if \( \tilde{q}_t \) is not constant, Proposition 3.8 applies, as \( x \nmid \tilde{q}_t \), which implies that \( \ell_{t+1} \) admits a factor of the form \( M^{r-1} u \) such that \( \tilde{q}_t | G \tilde{u} \). By Lemma 3.10, step 2 computes a polynomial \( \tilde{u} \) such that \( M^{r-1} u | M^{r-1} \tilde{u} \). It follows by Lemma 3.1(c) that \( u | \tilde{u} \), next that \( \tilde{q}_t | G \tilde{u} \), so that \( q \) divides \( q^* \), again.

Let us turn to the complexity analysis. Applying \( M \) to a polynomial requires no arithmetic operation. Each execution of step 1b amounts to \( b^r - 1 \) gcds of polynomials of degree less than or equal to \( d/b^r \), for a total cost of \( O(M(d) \log d) \) ops. The same argument applies to step 2. Similarly, the chain of lcms at step 1c requires
\[
O \left( \sum_{i=0}^{r-1} M(b^i \deg u_k) \log(b^i \deg u_k) \right) = O(M(d) \log d) \text{ ops},
\]
as \( \sum_{i=0}^{r-1} b^i \deg u_k = O(d) \). Since there are at most \( \rho^{-1} (1 + \deg \ell_r) \) iterations of steps 1b and 1c, the cost of step 1 is \( O(\rho^{-1} (\deg \ell_r) M(d) \log d) \). If \( b = 2 \), then \( \rho = 1 \) and the cost of step 1 is \( O((\deg \ell_r) M(d) \log d) \). If \( b \geq 3 \), then \( \rho \geq b^{-1} \) and the cost is \( O(b^{-r} (\deg \ell_r) M(d) \log d) \).

The computation of \( G \tilde{u} \) from \( \tilde{u} \) at step 3 can be performed in \( O(M(bd)) \) ops \([10, 3]\) and the final product can be computed in \( O(M(d) \log d) \) ops using a product tree.

Proposition 3.13 implicitly provides a bound on \( \deg q \) that essentially (when \( \bar{v} = 0 \) and \( \bar{u} = 1 \), exactly) matches that of Bell and Coons [2, Proposition 2]. However, a tighter bound holds, especially for \( b = 2 \).

**Proposition 3.14.** With the notation above, \( q \) has degree at most \( 3 \deg \ell_r / b^r \).

**Proof.** Let \( g = M^r q \land T q \). On the one hand, (3.2) implies \( M^r q | \ell_r g \), so that \( b^r \deg q \leq \deg \ell_r + \deg g \). On the other hand, \( M q \) divides \( h = M^r q \lor T q \) by definition of \( T \), hence \( g M q \) divides \( gh = (M^r q) (T q) \), whence
\[
(b + 1) \deg g \leq \frac{b^{r+1} - 1}{b - 1} \deg q.
\]
Comparing the two inequalities leads to
\[
\deg q \leq \frac{(b^2 - 1) \deg \ell_r}{b^{r+2} - b^{r+1} - b^r + 1} \leq \frac{(b^2 - 1) \deg \ell_r}{b^r (b^2 - b - 1)} \leq \frac{3 \deg \ell_r}{b^r}
\]
since \( (b^2 - 1)/(b^2 - b - 1) \leq 3 \) for \( b \geq 2 \). \( \square \)
Remark 3.15. The previous discussion to find $q^*$ is entirely based on (3.1) in the case $j=r$ and on expressing the solution $q$ with a minimal denominator $x^5 q$. Noting that (3.1) actually holds also for $j \neq r$ and even if $p \wedge q \neq 1$, we may apply it with $0 \leq j \leq r-1$ to a potential solution written in the form $p/(x^5 q^*)$ to get additional constraints involving $\ell_0, \ldots, \ell_{r-1}$ that can be used to remove some factors from $q^*$.

3.4. An alternative bound. We now describe an alternative method for computing denominator bounds. While it yields coarser bounds, our estimate for its computational cost is better, so that it may be a superior choice in some cases. The results of this subsection are not used in the sequel.

Proposition 3.16. If $x^5 q \in \mathbb{K}[x]$ is the denominator of a rational solution of (EQN) written in lowest terms, then it holds that

$$q \mid (G^r \ell_r) (G^{r+1} \ell_r) \cdots (G^{r+K} \ell_r), \quad K = \lceil \log_b(3 \deg \ell_r) \rceil - r.$$

Proof. Suppose $f$ is monic irreducible and $m$ is positive such that $f^m \mid q$, and consider the condition

$$M^{r+j} f \mid \bigvee_{i=0}^r M^i q.$$  \hfill (3.5)

Clearly, (3.5) is satisfied for $j = 0$, while it requires

$$b^{r+j} \deg f \leq \frac{b^{r+1} - 1}{b - 1} \deg q,$$

which in turn implies $j \leq \log_b \deg q$. Plugging in the bound from Proposition 3.14, we obtain $j \leq \log_b(3 \deg \ell_r) - r$.

Choose $j$ maximal such that (3.5) holds. Then $M^{r+j} f$ cannot divide $T q$, and by Lemma 3.3, $M^{r+j} f$ is squarefree. Let $h$ be a monic irreducible factor of $M^{r+j} f$ not dividing $T q$. In the rest of the proof, we write $\text{sqrfree} p$ for the squarefree part of any polynomial $p$. For all $k \geq 0$, set $h_k = \text{sqrfree}(G^k h)$, and denote by $m_k$ the multiplicity of $h_k$ as a factor of $T q$. Thus $m_0$ is zero by definition of $h$. Continuing with $k \geq 0$, Lemma 3.1(b) implies $G^k h \mid MG^{k+1} h$, so that $h_k \mid \text{sqrfree}(MG^{k+1} h) \mid M \text{sqrfree}(G^{k+1} h) = M h_{k+1}$. As $h_k^{m_k \! \mid \! T q}$, we deduce that $h_k^m \mid M h_{k+1}^m \mid MT q \mid T q \wedge M^{r} q$, then, by using (3.2), $h_k^{m_k+1} \mid \ell_r T q$. The definition of $m_k$ then yields $h_k^{b^{r+j} \delta_k \ell_r}$ for $\delta_k = \max(m_{k+1} - m_k, 0)$.

Now, restrict $k$ to the interval $j < k \leq r + j$. Then, by Lemma 3.1(b),

$$h_k \mid G^k h \mid G^k M^{r+j} f = (M^{r+j-k} f)^{b^k}, \hfill (3.6)$$

and as $h_k$ is squarefree, $h_k \mid M^{r+j-k} f$. Since $f^m \mid q$ and $0 \leq r + j - k < r$, $h_k^m$ divides $T q$, implying $m_k \geq m$.

By Lemma 3.1(b) and Equation (3.6), $G^{r+j-k} h_k$ is $f^{b^{r+j} j}$, so that $f$ divides the former. Then,

$$f^{b^{r+j} \delta_k} \mid G^{r+j-k} h_k^{b^{r+j} \delta_k} \mid G^{r+j-k} \ell_r.$$

Forming the product of these bounds for $k$ ranging from 0 to $j$, we get $f^m \mid \prod_{k=0}^j G^k \ell_r$, as $m \leq m_{j+1}$ and $m_0 = 0$. The result follows by considering all possible $(f, m)$ such that $f^m \mid q$. \hfill \qed

Proposition 3.17. One can compute a polynomial $q^* \in \mathbb{K}[x]$ of degree at most $d (\log_b d - r + 2)$ and such that $q \mid q^*$ in $O(M \log d \log d)$ ops.
The product then takes proof. Define Proposition 3.19. combined with Lemma 2.5, it has degree at most Proposition 3.13 combined with Lemma 2.2, rational solution of \{ equation. Assuming \( p \in \mathbb{K}[x] \), not necessarily coprime, satisfy \( L(p/q) = 0 \), then \( \deg p \) is at most \( \deg q + \lfloor d/(b^r - b^{r-1}) \rfloor \).

The procedure to obtain rational solutions is summarized in Algorithm 9.

**Algorithm 9. Rational solutions**

\[
\text{Input: A linear Mahler equation of the form (EQN).}
\]
\[
\text{Output: A basis of its space of rational function solutions.}
\]

(1) Set \( \delta = \max \deg \ell_k \).
(2) If \( \delta < b^{-r-1} \): return \( \{1\} \) if \( L(1) = 0 \), and \( \emptyset \) otherwise.
(3) Compute \( q^* \) using Algorithm 8. Set \( v = \lfloor \delta/(b^r - b^{r-1}) \rfloor \).
(4) For \( 0 \leq k \leq r \), set \( e_k = \lfloor bδ/(b-1) \rfloor - b^k v \) and
\[
\hat{\ell}_k = x^{e_k} \ell_k \prod_{0 \leq i \leq r, i \neq k} \hat{M}^i q^*.
\]
Set \( \hat{L} = \hat{\ell}_r M^r + \cdots + \hat{\ell}_0 \).
(5) Call Algorithm 5 on the equation \( \hat{L}p = 0 \), with \( w = \deg q^* + 2v + 1 \), to compute a basis \( \langle p_1, \ldots, p_\sigma \rangle \) of its polynomial solutions of degree less than \( w \).
(6) Return \( \langle p_k/(x^{\hat{v}} q^*) \rangle \) for \( 1 \leq k \leq \sigma \).

**Proof.** If \( \deg \ell_r < b^{-r-1} \), return 1. This is a valid bound by Corollary 3.9. Otherwise, return the bound from Proposition 3.16. As with the previous bound, the \( G^k \ell_r \) up to \( k = r + K = O(\log d) \) can be computed for a total of \( O(M(bd) \log d) \) ops [10, 3]. The product then takes \( O(M(d \log d) \log d) \) ops. \( \square \)

### 3.5. Computing numerators.
In order to obtain a basis of rational solutions \( y \) of (EQN), it suffices to obtain a bound \( x^{\hat{v}} q^* \) on denominators as in §3.3, to construct an auxiliary equation corresponding to the change of unknown functions \( y = \hat{y}/(x^{\hat{v}} q^*) \), and to search for its polynomial solutions \( \hat{y} \). We first note the following consequence of Lemma 2.5, already proved by Bell and Coons [2, Prop. 2].

**Proposition 3.18.** If \( p, q \in \mathbb{K}[x] \), not necessarily coprime, satisfy \( L(p/q) = 0 \), then \( \deg p \) is at most \( \deg q + \lfloor d/(b^r - b^{r-1}) \rfloor \).

The procedure to obtain rational solutions is summarized in Algorithm 9.

**Proposition 3.19.** Algorithm 9 computes a basis of rational solutions of its input equation. Assuming \( d \geq b^{-r-1} \), it runs in \( O(d M(d) + 2r d^2 + M(2^r d)) \) ops when \( b = 2 \) and \( O(b^{-r} d M(d)) \) ops when \( b \geq 3 \). Assuming further \( M(n) = O(n) \), it runs in \( O(2r d^2) = O(d^3) \) ops when \( b = 2 \) and in \( O(b^{-r} d^2) \) ops when \( b \geq 3 \).

**Proof.** Define \( \delta \) as in step 1, so that \( \delta \leq d \). If \( \delta < b^{-r-1} \), the algorithm will stop after step 2. In this case, Corollary 3.9 states that there are no nonconstant rational solution. Therefore, the vector space of rational solutions is \( \mathbb{K} \) when \( L(1) = 0 \) and \( \{0\} \) otherwise.

Otherwise, the algorithm continues with \( d \geq b^{-r-1} \). Assume that \( y \in \mathbb{K}(x) \) is a rational solution of \( L \hat{y} = 0 \), and let \( p = x^{\hat{v}} q^* y \) for \( q^* \) and \( \hat{v} \) computed as in step 3. By Proposition 3.13 combined with Lemma 2.2, \( p \) is a polynomial. By Proposition 3.18 combined with Lemma 2.5, it has degree at most \( \deg (x^{\hat{v}} q^*) + \hat{v} = \deg q^* + 2\hat{v} \). Plugging \( y = p/(x^{\hat{v}} q^*) \) into \( L \hat{y} = 0 \) and multiplying the resulting equation by the polynomial \( x^{[b\hat{v}/(b-1)]} \prod_{i=0}^{r} M^i q^* \), we see that \( p \) satisfies \( \hat{L} p = 0 \), where \( \hat{L} \) is defined
as in step 4. As $b^k v \leq b^\delta/(b - 1)$ for $k \leq r$, the $e_k$ are nonnegative and the $\tilde{\ell}_k$ are polynomials. Thus Algorithm 5 applies and, by Proposition 2.14, $p$ belongs to the span of the $p_k$ computed at step 5 of Algorithm 9. Conversely, for all $k$, the fraction $p_k/(x^b p^*)$ is a solution of $L y = 0$.

After step 2, we have $b^r = O(d)$, that is, $r = \tilde{O}(1)$. By Proposition 3.13, the cost of step 3 is $O(d M(d))$ ops when $b = 2$ and $O(b^{-r}d M(d))$ ops when $b \geq 3$. Define

$$\tilde{d} = \frac{2b - 1}{b - 1} d + \frac{b^{r+1} - 1}{b - 1} \deg q^* = \begin{cases} O(2^r d), & b = 2, \\ O(d), & b \geq 3, \end{cases}$$

where the asymptotic bounds follow from Proposition 3.13. Each polynomial $\tilde{\ell}_k$ defined at step 4 then satisfies

$$\deg \tilde{\ell}_k \leq e_k + \delta + \frac{b^{r+1} - 1}{b - 1} \deg q^* \leq \tilde{d}$$

and is the product of $r + 1$ polynomials, so their computation takes $O(r M(\tilde{d})) = \tilde{O}(M(\tilde{d}))$ ops. Observe as well that $1 \leq w = O(\tilde{d}/b^r)$. According to Proposition 2.14, step 5 thus requires $\tilde{O}(b^{-r}d^2 + M(\tilde{d}))$ ops, which dominates the cost of step 4. Taking the bounds (3.7) into account, we get that step 5 is dominated by step 3 when $b \geq 3$, so that the total cost is $\tilde{O}(d M(d) + 2^r d^2 + M(2^r d))$ ops when $b = 2$ and $\tilde{O}(b^{-r}d M(d))$ ops when $b \geq 3$. With fast multiplication, $M(n) = \tilde{O}(n)$, this simplifies to the announced complexity estimates.

**Example 3.20.** We continue Example 3.12. We have seen that the denominator bound is $q^* = (2x - 1)(x^2 - x - 1)(8x - 1)(x^2 - 4x - 1)$. We set $\tilde{y} = q^* y$, so that $L y = 0$ if and only if $L \tilde{y} = 0$, where $\tilde{L} = \tilde{L}_2 M^2 + \tilde{L}_1 M + \tilde{L}_0$ for

\[
\tilde{L}_2 = (2x - 1)(8x - 1)(x^2 - x - 1)(x^2 - 4x - 1) \times (4x^2 + 2x + 1)(2x^8 - 3x^3 - x + 3)(x^4 + x^3 + 2x^2 - x + 1),
\]

\[
\tilde{L}_1 = -(8x - 1)(x^2 + 1)(x^2 - 4x - 1)(2x^3 - 1)(x^4 + 1) \times (x^6 - x^3 - 1)(4x^6 + 2x^3 + 1)(2x^{10} - x^9 - x + 3)(x^{12} + x^9 + 2x^6 - x^3 + 1),
\]

\[
\tilde{L}_0 = x^2(2x - 1)(2x^2 + x + 1)(x^2 - x + 1)(x^2 - x - 1) \times (4x^2 + 2x + 1)(2x^3 - 1)(x^4 + x^3 + 2x^2 - x + 1) \times (x^6 - x^3 - 1)(4x^6 + 2x^3 + 1)(x^{12} + x^9 + 2x^6 - x^3 + 1)(2x^{12} - x^9 - x^3 + 3).
\]

We have to compute the complete set of polynomial solutions of $L \tilde{y} = 0$. The degree of $\tilde{L}_2, \tilde{L}_1, \tilde{L}_0$ are respectively 16, 46, 54. Using Lemma 2.5, we find that the degree of a nonzero polynomial solution is necessarily 4 or 5. Following Algorithm 6, we equate the coefficients on both sides of $L \tilde{y} = 0$ up to degree 54, and we obtain that $\tilde{y} = \tilde{y}_0 + \cdots + x^5 \tilde{y}_5$ is solution of $L \tilde{y} = 0$ if and only if the vector $(\tilde{y}_0, \ldots, \tilde{y}_5)$ is solution of a system of $h = 163$ equations. A basis of solutions turns out to consist of $(2x - 1)(8x - 1)(x^2 - 4x - 1)$ and $(x^2 - x - 1)(8x - 1)(x^2 - 4x - 1)$. Consequently, a basis of rational-function solutions of $L y = 0$ consists of

$$\frac{1}{2x - 1} \text{ and } \frac{1}{x^2 - x - 1}.$$
Input: A linear Mahler operator $L$ with coefficients in $\mathbb{K}[x]$.

Output: A set of linear Mahler operators with coefficients in $\mathbb{K}[x]$ and $M$-valuation zero.

1. If $L = 0$, return $\emptyset$.
2. If $L$ has $M$-valuation 0, return $\{L\}$.
3. Return the union of the results of calling the algorithm recursively on each section $S_i(L)$ for $0 \leq i < b$.

Algorithm 10. Split of (opr).

Algorithm 9 with Proposition 2.19 about possible ramifications, we obtain an algorithm that computes a basis of solutions of (eqn) in $\mathbb{K}((x))$. Assuming $M(n) = \tilde{O}(n)$, it runs in $\tilde{O}(2^{3r}d^3)$ ops when $b = 2$ and in $\tilde{O}(b^r d^2)$ ops when $b \geq 3$.

Note that, as in §2.7, these complexity bounds hold even if $\ell_0$ is zero.

4. The case $\ell_0 = 0$ and an algorithm for computing gcrd’s

In this section, we drop the assumption $\ell_0 \neq 0$. More precisely, we consider a linear Mahler equation of the form (eqn), with $\ell_0 = \cdots = \ell_{w-1} = 0$ and $\ell_r \ell_w \neq 0$. We call the integer $w$ the $M$-valuation of (eqn) and $d = \max_{k=w,\ldots,r} \deg \ell_k$ its degree. We define the $M$-valuation and the degree of the corresponding operator (opr) similarly. The goal of this section is to compute a linear Mahler equation with $M$-valuation equal to 0, such that the new equation and (eqn) have the same set of series solutions in $\mathbb{K}((x))$.

The algorithm proposed here, Algorithm 11, can be seen as an improvement over an algorithm given by Dumas in his thesis [8, §3.2.1]. In particular, Algorithm 10, borrowed from [8], performs the subtask of splitting an operator of positive $M$-valuation into a system of operators of zero $M$-valuation while preserving the solution set in $\mathbb{K}((x))$. Dumas’s algorithm next makes use of the right Euclidean structure of the algebra $M(\mathbb{K})$ of linear Mahler operators with coefficients in $\mathbb{K}(x)$, and transforms the system into a single, equivalent equation by computing a gcrd (greatest common right divisor) via Euclidean divisions. The problem of this approach is that the degree of the obtained equation explodes in the process. To avoid this, we change the second step of algorithm in [8] so as to reuse Algorithm 10 and cancellations of trailing instead of leading coefficients.

The splitting process of Algorithm 10 is explained in terms of section maps $S_i$, each of which maps a polynomial in $x$ and $M$ to a polynomial in $x$ and $M$, and whose collection plays the role of a partial inverse for $M$: for $0 \leq i < b$, let $S_i$ be the $K$-linear map that sends $x^j M^{k+1}$ to $x^{(j-i)/b} M^k$ if $(j-i)/b$ is an integer and to 0 otherwise.

Lemma 4.1. Let $L$ be a linear Mahler operator $L$ of the form (opr) and have degree $d$ and positive $M$-valuation. Then, whenever $0 \leq i < b$, the section $S_i(L)$ has degree at most $d/b$. Additionally, $L$ can be reconstructed from its sections by

$$L = \sum_{i=0}^{b-1} x^i M S_i(L).$$
We call this operator the Mahler operators of order at most $d$.

Algorithm 11 does not have the same set of solutions in $\mathbb{K}$ as in [8], we use in Algorithm 11 linear combinations of $\mathbb{K}$ that kill constant terms: given two nonzero Mahler operators $L_1$ and $L_2$ with coefficients in $\mathbb{K}$, $M$-valuation zero, and coefficient of $M^0$ respectively $c_1$ and $c_2$, we write $R(L_1, L_2)$ for the operator $c_2 L_1 - c_1 L_2$, whose coefficient of $M^0$ is zero. We call this operator the interreduction of $L_1$ and $L_2$ and a step of the algorithm that replaces an operator $L_1$ by an interreduction $R(L_1, L_2)$ a reduction step.

**Lemma 4.2.** Let $L$ be a linear Mahler operator of the form (OPR), with order $r$, $M$-valuation $w$, and degree $d$. Then, Algorithm 10 returns a set of nonzero linear Mahler operators of order at most $r - w$, $M$-valuation 0, and degree at most $d b^{-w}$.

**Proof.** This is shown by a straightforward induction on $w$. □

Instead of considering usual Euclidean divisions according to decreasing powers, which would compute a gcrd as in [8], we use in Algorithm 11 linear combinations that kill constant terms: given two nonzero Mahler operators $L_1$ and $L_2$ with coefficients in $\mathbb{K}[x]$, $M$-valuation zero, and coefficient of $M^0$ respectively $c_1$ and $c_2$, we write $R(L_1, L_2)$ for the operator $c_2 L_1 - c_1 L_2$, whose coefficient of $M^0$ is zero.

**Lemma 4.3.** Let $L$ be a system of Mahler operators. Replacing an element $L$ of $L$ by its sections $S_0(L), \ldots, S_b(L)$ does not change the set of solutions of $L$ in $\mathbb{K}((x))$. Nor does replacing $L_1$ by the interreduction $R(L_1, L_2)$ where $L_1, L_2$ are distinct elements of $L$.

**Proof.** The second claim is obvious. Regarding the first one (already in [8, §3.2.1]), the decomposition (4.1) shows that any common solution of the $S_i(L)$ is a solution of $L$. If, conversely, $y$ is an unramified solution of $L$, then the $x^i MS_i(L) y$, $0 \leq i < b$, have disjoint support, hence $S_i(L) y = 0$ for all $i$. □

Here, the degree of $R(L_1, L_2)$ may well be the sum of the degrees of $L_1$ and $L_2$, but having generated a multiple of $M$ makes it possible to apply splitting and keep degrees under control. This leads to Algorithm 11, whose correctness and complexity are given in the following proposition.

It is worth mentioning that, in general, the equation $\tilde{L}(y) = 0$ returned by Algorithm 11 does not have the same set of solutions in $\mathbb{K}((x^{1/\alpha}))$ as the equation $L(y) = 0$. As an example, let $b = 2$ and consider $L = M^2 - xM$. We have $\tilde{L} = 1$, and the solution space in $\mathbb{K}((x^{1/\alpha}))$ of $\tilde{L}(y) = 0$ is $\{0\}$. On the other hand, the solution space in $\mathbb{K}((x^{1/\alpha}))$ of $L(y) = 0$ is the $\mathbb{K}$-vector space spanned by $x^{1/2}$.
Proposition 4.4. The operator $L$ has the same set of solutions in $\mathbb{K}((x))$ as the operator $\tilde{L}$ returned by Algorithm 11. This operator has order $\tilde{r} \leq r - w$, $M$-valuation 0, and degree $\tilde{d} \leq db^{-w}$. Furthermore, Algorithm 11 runs in $O(rb^w M(d/b^w))$ ops.

Proof. Because $L \neq 0$ and by construction of Algorithm 10, the initial set $\mathcal{L}$ is nonempty. Next, by construction of Algorithm 11, at any time of a run, $\mathcal{L}$ is nonempty and contains only elements of outputs from Algorithm 10, so that, by Lemma 4.2, if Algorithm 11 terminates, its output must be nonzero and of $M$-valuation zero. Lemma 4.3 implies that the original operator $L$, the system $\mathcal{L}$ at any time of the run, and therefore the final operator $\tilde{L}$, all share the same set of solutions in $\mathbb{K}((x))$.

Let us prove the bound on the order and the degree of $\tilde{L}$. By Lemma 4.2, the set $\mathcal{L}$ computed at step 1 consists of Mahler operators with orders bounded by $r - w$ and degrees bounded by $db^{-w}$. These bounds keep on holding after each run of the loop body at step 2: As the operators $L_1$ and $L_2$ chosen at step 2a satisfy the property, their combination $R(L_1, L_2)$ (including the case it is zero) has order bounded by $r - w$, degree bounded by $2db^{-w}$, and positive valuation. By Lemma 4.2, the set $\mathcal{L}'$ computed at step 2b consists of Mahler operators with orders bounded by $r - w - 1$ and degrees bounded by $2db^{-(w+1)}$. As $2/b \leq 1$, the set $\mathcal{L}$ retains the property after the update at step 2c. Therefore, if the algorithm terminates, it returns at step 3 an element of $\mathcal{L}$, therefore with the announced order and degree bounds.

We finally prove termination and complexity by a joint argument. To this end, we represent the process of Algorithm 11 by an oriented tree labeled by operators $L^n_w$, for integers $n$ and words $w$ on the alphabet $\{0, \ldots, b-1\}$. These operators $L^n_w$ will be the operators considered during the execution of the algorithm. This tree is rooted at the node labeled $L^n_0 = L$, and evolves by following the execution of Algorithm 11. Each time a section of an operator $L^n_w$ is computed by the subtask of Algorithm 10, whether it be at step 1 or at step 2b, the tree is augmented by new edges from $L^n_w$ to its subsection $L^n_{wx} = S_j(L^n_w)$. For each choice of $L_1 = L^n_w$ and $L_2 = L^{n'}_w$ at step 2a, the tree is augmented by a new edge labeled $L^n_{wx}$, from $L^n_w$ to $L^{m+1}_w$, if $m$ is the larger upper index in the tree before reduction. Thus, one obtains that at each stage of the execution, the set $\mathcal{L}$ is equal to the collection of nonzero leaves of the current tree. Now, by construction of the tree and by design of the algorithm, a reduction step results either in a zero operator or in an operator with positive $M$-valuation that is immediately split to its sections. Therefore, following a path from the root to a leaf, two reduction edges can only appear if separated by at least one section edge. As section edges reduce orders by at least 1, while reduction edges do not increase orders, the tree has to be finite and the algorithm terminates. The only arithmetic operations of the algorithm are the polynomial products involved in the computation of the $R(L_1, L_2)$ at step 2b. It was proved above that any operator of $\mathcal{L}$ has degree bounded by $d/b^w$. Because operators all have order at most $r$ and as the size of the tree bounds the number of reductions, the algorithm has total complexity $O(rb^w M(d/b^w))$. \hfill \square

Remark 4.5. A slightly better complexity can be obtained by a variant of Algorithm 11, in which the $L_1$ at step 2a is not chosen as having maximal order, but according to a notion of depth in the tree introduced for the proof of Proposition 4.4.
with
\[
L_1 = x^9(1 - x^{15} + x^{51} + x^{54} - x^{87} + x^{108})(1 - x^{12} + x^{24}),
\]
\[
L_2 = -x^3(1 + x^6 - x^{20} - x^{21} + x^{30} + x^{32} + x^{33} + x^{36} - x^{44} - x^{45} + x^{54} + x^{56}
+ x^{57} + x^{60} - x^{68} + x^{80} + x^{81} + x^{84} + x^{90} - x^{92} - x^{93} + x^{104}
+ x^{105} + x^{108} + x^{114} - x^{116} - x^{117} + x^{138} + x^{144}),
\]

Doing so guarantees a better behavior of degrees, with a geometric decrease with depth, as opposed to the uniform bound \(d/b^w\) used in the proof above.

Define the depth \(\beta\) of a node \(L_w^n\) in the tree as the number of section edges from the root \(L_0^0\) to \(L_w^n\), and change the strategy at step 2a to choose \(L_1\) among the elements of \(L\) of lowest depth. By another induction, \(L_w^n\) has order not more than \(r - \beta\), as in the proof above, but its degree is not more than \(d/b^w\) if \(\beta \leq w\), and not more than \((2/b)\beta^{3-w}(d/b^w)\) if \(\beta > w\). A bound on the complexity becomes
\[
\sum_{\beta=w}^{r} (r+1)b^\beta M \left( \frac{2\beta^{3-w}d}{b^\beta} \right) \leq O(r M(2^r d/b^w)).
\]

This bound is better than the original complexity \(O(rb^r M(d/b^w))\) when \(b \geq 3\).

For \(b = 2\), the new bound is not tight and the variant algorithm has the same complexity bound as Algorithm 11.

**Example 4.6.** We apply Algorithm 11 with \(b = 3\) and the operator
\[
L = \ell_1 M + \ell_2 M^2 + \ell_3 M^3 + \ell_4 M^4
\]
with
\[
\ell_1 = x^9(1 - x^{15} + x^{51} + x^{54} - x^{87} + x^{108})(1 - x^{12} + x^{24}),
\]
\[
\ell_2 = -x^3(1 + x^6 - x^{20} - x^{21} + x^{30} + x^{32} + x^{33} + x^{36} - x^{44} - x^{45} + x^{54} + x^{56}
+ x^{57} + x^{60} - x^{68} + x^{80} + x^{81} + x^{84} + x^{90} - x^{92} - x^{93} + x^{104}
+ x^{105} + x^{108} + x^{114} - x^{116} - x^{117} + x^{138} + x^{144}),
\]
\[ \ell_3 = (1 + x^3 - x^5 + x^{17} + x^{18} + x^{21} - x^{23} - x^{29} + x^{35} + x^{36} + x^{39} - x^{47} + x^{54} + x^{57} + x^{72} + x^{75} + x^{90} + x^{93} - x^{95} + x^{107} + x^{108} + x^{111} - x^{113} - x^{119} + x^{125} + x^{126} + x^{129} - x^{137} + x^{144} + x^{147}), \]
\[ \ell_4 = -(1 + x^{27} + x^{54})(1 - x^{27} + x^{54})(1 - x^5 + x^{17} + x^{18} - x^{29} + x^{36}). \]

Starting from \( L'_0 = L \), we compute its sections (see Fig. 6, blue edges): first, \( L_0^0 = S_0(L'_0) \), which has \( M \)-valuation 0 so that the process of splitting stops for it; next, \( L_1^0 = S_1(L_0^0) \), which is zero and is dropped; last, \( L_2^0 = S_2(L'_0) \), which has \( M \)-valuation 1. Splitting continues for the latter and provides \( L_{20}^0 = S_0(L_1^0) \), \( L_{21}^0 = S_1(L_0^0) \), \( L_{22}^0 = S_2(L_0^0) \), all with \( M \)-valuation 0. Note that during this splitting, the operators \( L_0^0, L_1^0, \) and \( L_2^0 \) disappear. A reduction is made (see Fig. 6, red edges) where \( R(L_0^0, L_{21}^0) = L_6^e \) replaces \( L_0^0 \). The process continues and, at the end, there only remains

\[ L_1 = x^5(1 + x + x^2)(1 - x + x^2)(1 - x^4 + x^6) \]
\[ - x^3(1 + x + x^2)(1 - x + x^2)(1 - x^2 + x^4 - x^6 + x^8)(1 + 2x^2 + x^4)M \]
\[ + x^3(1 + x + x^2)(1 - x + x^2)(1 + x^3 + x^6)(1 - x^3 + x^6)M^2. \]

It is worth noting that \( L_1 \) has a content \( c = x^3(1 + x + x^2)(1 - x + x^2) \), so that we can write \( L_1 = cL_1^1 \) where \( L_1^1 \) is a primitive polynomial (with respect to \( M \)). The computation shows that \( L \) is in the left ideal generated by \( L_1^1 \) in the algebra \( M(\mathbb{Q}) \). This and exhibiting the \( M \)-valuation \( w = 1 \) of \( L \) provides factorizations \( L = L'M = L''M'L_1^1 \). We can say that \( M^w \) has been pushed as much as possible to the left. Using Algorithm 9, we find that a basis of solutions of \( L_1^1 \) in \( \mathbb{K}(x) \) is given by 1 and \( x^{-1} \). Since \( L_1^1 \) has order two, this also forms a basis of solutions of \( L_1^1 \) in \( \mathbb{K}(\!(x)\!) \), as a consequence of Proposition 2.3, and by Proposition 4.4, a basis of solutions of \( L \) in \( \mathbb{K}(\!(x)\!) \).

We now proceed to prove that Algorithm 11 indeed computes a gcrd with controlled degree. This is proved in Theorem 4.9 below, using the following lemmas.

**Lemma 4.7.** For any operators \( P_1, P_2 \), and any integer \( i \) such that \( 0 \leq i < b \), \( S_i(P_1MP_2) = S_i(P_1M)P_2 \).

**Proof.** By linearity, it is sufficient to consider \( P_1 = x^{j_1}M^{k_1} \) and \( P_2 = x^{j_2}M^{k_2} \). Then, \( P_1MP_2 = x^{j_1+kb+1}j_2M^{k_1+k_2+1} \). Either \( b \) divides \( j_1 - i \) and
\[ S_i(P_1MP_2) = x^{(j_1-i)/b+jb+j_2M^{k_1+k_2+1} = x^{(j_1-i)/b}M^{k_1}x^{j_2}M^{k_2} = S_i(P_1M)P_2, \]
or \( b \) does not divide \( j_1 - i \) and both extreme terms are zero, thus equal again. \( \square \)

**Lemma 4.8.** For any operators \( P_1, P_2, \) and \( P \), all of \( M \)-valuation 0, let \( c \) be the coefficient of \( M^0 \) in \( P \). Then, \( R(P_1P, P_2P) = cR(P_1, P_2) \).

**Proof.** The property holds, as obviously the coefficient of \( M^0 \) in a product is the product of the coefficients of \( M^0 \) in the factors. \( \square \)

**Theorem 4.9.** Steps 2 and 3 of Algorithm 11 compute a gcrd of the elements of the split \( L \) of \( L \) obtained at step 1. The degree of this particular gcrd is bounded by the maximal degree of the elements of \( L \).
Proof. Let $I$ denote the left ideal $\mathcal{M}(\mathbb{K})\mathcal{L}$ generated by $\mathcal{L}$ at any time in the run of the algorithm. Call $G$ the monic gerd of the elements of the set $\mathcal{L}$ as obtained from $L$ at the end of step 1. By (4.1), $G$ is a right factor of $L$. By the definition of $\mathcal{R}(\cdot, \cdot)$ and because of (4.1) again, the ideal $I$ can only increase during the run of the algorithm, so during step 2, $\mathcal{M}(\mathbb{K})L \subset \mathcal{M}(\mathbb{K})G \subset I$.

We show by induction that $G$ is a right factor of all elements of $\mathcal{L}$ at any time in step 2, in other words, that $I \subset \mathcal{M}(\mathbb{K})G$. This is true by the definition of $G$ when entering the loop. The set $\mathcal{L}$ contains only elements with $M$-valuation 0, and it cannot be empty when entering the loop, so $G$ has $M$-valuation 0 as well. At any step 2b, divisibility on the right by $G$ is preserved for $\mathcal{R}(L_1, L_2)$, by Lemma 4.8. As $\mathcal{R}(L_1, L_2)$ has positive $M$-valuation, one can choose $P_2 = G$ and find $P_1$ so as to write $\mathcal{R}(L_1, L_2) = P_1MP_2$. By Lemma 4.7, it follows that divisibility on the right by $G$ is also preserved for each element of $\mathcal{L}$, then for each element of the next value of $\mathcal{L}$.

As a consequence, during step 2, $I$ constantly equals $\mathcal{M}(\mathbb{K})G$. In particular, the final operator $\tilde{L}$ is proportional to $G$.

The degree bound was proved as part of Proposition 4.4. □

Remark 4.10. Note that the origin of the initial $\mathcal{L}$ as a split of $L$, at step 1 of Algorithm 11 plays no role in the proof of Theorem 4.9. Thus, Algorithm 11 implicitly contains an algorithm for computing the gerd of any family $\mathcal{L}$ of operators of $M$-valuation zero.

Remark 4.11. We developed Algorithm 11 without targeting a gerd and realized Theorem 4.9 (and Remark 4.10) only a posteriori. As Algorithm 11 indeed works by computing a gerd as the original algorithm in [8], it is now instructive to compare the result of Proposition 4.4 with bounds on the size of gcds of Mahler operators given by existing methods. Such a bound can be computed using a variant of the subresultant argument given by Grigor’ev [9, §5] in the differential case.

Let $L_1, \ldots, L_n$ be operators of respective order $r_1 \geq r_2 \geq \cdots \geq r_n \geq 1$ and degree $d_1, \ldots, d_n \leq \delta$. Let $G = U_1L_1 + \cdots + U_nL_n$ be their greatest common right divisor. We can assume that the order of each term $U_iL_i$ is less than $t = r_1 + r_n$. Indeed, for all $i, j$ the linear equation $V_{i,j}L_i = V_{j,i}L_j$ with $V_{i,j}$, resp. $V_{j,i}$, constrained to have degree at most $r_j$, resp. at most $r_i$, has nontrivial solutions. Via Euclidean divisions $U_i = Q_iV_{i,n} + R_i$, we obtain $G = \sum_i(Q_iV_{i,n} + R_i)L_i = \sum_i Q_iV_{n,i}L_n + \sum_i R_iL_i = \sum_i \tilde{U}_iL_i$ where the $\tilde{U}_i$ for $i \leq n - 1$ have order less than $r_n$. The $n - 1$ first terms $U_iL_i$ as well as $G$ itself have order less than $r_1 + r_n$, hence the same must be true of $\tilde{U}_nL_n$.

Consider a Sylvester-like matrix $S \in \mathbb{K}[x]^{s \times t}$ with rows

$$\mathcal{R}(L_1), \mathcal{R}(ML_1), \ldots, \mathcal{R}(M^{t-1}L_1), \ldots, \mathcal{R}(L_n), \mathcal{R}(ML_n), \ldots, \mathcal{R}(M^{t-1}L_n),$$

where, for any operator $L = \sum_k \ell_kM^k$, we denote $\mathcal{R}(L) = (\ell_{t-1}, \ldots, \ell_0)$. Call $C_0, C_1, \ldots, C_{t-1}$ the columns of $S$, listed from right to left (so that $C_j$ contains the coefficients of $M^j$ in $M^kL_i$), and $C_{j,0}, C_{j,1}, \ldots, C_{j,s-1}$ the entries of $C_j$. Let $m$ denote the order of $G$, and choose $J \subseteq \{m + 1, \ldots, t - 1\}$ of cardinality $|J| = \text{rk} S - 1$ in such a way that the columns $C_j$ with $j \in J$ form a basis of the span of $C_{m+1}, \ldots, C_{t-1}$, while the $C_j$ for $j \in \{m\} \cup J$ form a basis of the full column space of $S$. To see that such a $J$ exists, consider a row echelon form of $S$: since $\mathcal{R}(G)$ belongs to the left image of $S$ and $G$ has minimal order among the nonzero elements of the ideal $\sum_i \mathcal{M}(\mathbb{K})L_i$, the rightmost pivot lies on column $m$. Further, let $I \subseteq \{0, \ldots, s - 1\}$ be such that the submatrix $(C_{j,i})$, $i \in I$, $j \in J \cup \{m\}$ of $S$ is
Input: A finite family \( \{L_i\}_{i=1}^s \) of linear Mahler operators with polynomial coefficients, orders at most \( r \), minimal \( M \)-valuation \( w \), and degrees at most \( d \).

Output: A linear Mahler operator \( \tilde{L} \) of order \( \tilde{r} \leq r \), \( M \)-valuation \( w \), and degree \( \tilde{d} \leq d \).

1. Write each \( L_i \) in the form \( L_i M^w \), for a polynomial \( L_i' \) in \( x \) and \( M \).
2. Let \( L \) be the union of the results of applying Algorithm 10 to the \( L_i' \)'s.
3. While \( L \) has at least two elements:
   a. choose \( L_1 \) with highest order in \( L \), then \( L_2 \) from \( L \setminus \{L_1\} \);
   b. compute the result \( L' \) of applying Algorithm 10 to the interreduction \( R(L_1, L_2) \);
   c. replace \( L \) by \( (L \setminus \{L_1\}) \cup L' \).
4. Write \( \tilde{L} \) for the single element of the singleton \( L \) and return \( \tilde{L} M^w \).

Algorithm 12. Computation of a gcrd of an arbitrary family.

nonsingular. Call \( D_{mn} \) the corresponding minor, and more generally define \( D_k \) as the determinant of the submatrix \( (C_{j,i})_{i \in I, j \in J} \), extended on the right by a copy of \( C_k \). Expanding \( D_k \) along the last column yields \( D_k = \sum_{i=0}^{n} u_i C_{k,i} \), where the \( u_i \) do not depend on \( k \). For each \( k > m \), the determinant \( D_k \) is zero, as \( C_k \) is in the span of the \( C_j \) for \( j \in J \). It follows that the vector \((0, \ldots, 0, D_m, \ldots, D_0)\) belongs to the left image of \( S \). Thus, there is a gcrd of \( L_1, \ldots, L_n \) with polynomial coefficients whose coefficients are minors of \( S \).

The entries of \( S \) have degree bounded by \( \delta' = \max_{i=1}^n (b^r - 1 d_i) \). Therefore, the degree of \( G \) is at most \( ts' \leq 2r_1 b^{r_1 - 1} \delta \leq r_1 b^{r_1} \delta \). Using fast polynomial linear algebra, it is plausible that one could actually compute \( G \) based on this approach with a complexity of the type \( \tilde{O}(\delta' t^s) = \tilde{O}(b^{r_1} \delta) \). Now, the gcrd in the algorithm of [8] is that of a family of iterated sections of the input operator \( L \). In terms of the order \( r \) and degree \( d \) of \( L \), this family can involve elements simultaneously of order \( r - 1 \) and degree \( d/2 \). Thus, Grigor’ev’s approach (at least in a straightforward way) would lead to a complexity bound similar to that of Proposition 4.4, but an exponentially worse bound on the degree of the output for large \( r \).

This result leaves open the question of devising algorithms for computing solutions of linear Mahler equations that run in polynomial time in \( r \) and \( d \), for all possible combinations of these parameters, even when the trailing coefficient \( \ell_0 \) of the equation is zero. In particular, it would be interesting to see if the bounds on the size of an operator equivalent to \( L \) implied by Algorithm 10 would be enough to extend the algorithms of §2–3 to the case where \( \ell_0 \) is zero, without going through the explicit computation of such an operator.

We end the section by providing an extension of Algorithm 11, which computes a gcrd for a family of operators of arbitrary \( M \)-valuations.

Theorem 4.12. Algorithm 12 computes a gcrd of the input operators \( L_1, \ldots, L_s \).
Proof. Observe that the minimal $M$-valuation of operators in a family is the minimal $M$-valuation of elements of the left ideal generated by the family, in particular, the $M$-valuation of any gcrd of the family. This justifies the general design of the algorithm, with the factorization of $M^w$ on the right at step 1.

By construction, the $L'_i$’s thus obtained have orders at most $r - w$ and degrees at most $d$, and at least one, say $L'_1$, has $M$-valuation zero. Let $G'$ denote the monic gcrd of the $L'_i$, which, as $L'_1$, has $M$-valuation zero. By Lemma 4.7, $G'$ is a right-hand factor of all elements of the set $L$ computed at step 2. By a proof similar to the one for Theorem 4.9, it remains so for all subsequent values of $L$, so for the $\tilde{L}$ of step 4 as well.

As $\tilde{L}$ is also obviously a right-hand factor of all previously computed operators, including the $L'_i$’s, $\tilde{L}$ is a gcrd of the latter. This concludes the proof. $\blacksquare$

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