Interpolation of Sparse Rational Functions Without Knowing Bounds on Exponents

by

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Abstract

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We present an algorithm for the (black box) interpolation of t-sparse rational functions without knowing bounds on exponents of their sparse representations.

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Introduction.

A t-sparse rational function is a function that can be written as a quotient of two polynomials, each containing at most t terms. We show in this paper that, if we are given a black box to evaluate a t-sparse rational function f with integer coefficients, then one can bound the exponents appearing in a t-sparse representation of f by making $2(t+1)^t - 1$ black box evaluations in the univariate case and $O(nt^t)$ black box evaluations in the *n*-variable case. Using this, we also give the first algorithm for interpolation of t-sparse rational functions without knowing bounds on exponents and show that for fixed t this problem is in polynomial parallel time (sequential storage). For the corresponding versions of bounded degree rational interpolation (where the bound on the degree is part of the input) see [S 73], [K 86], [KT 88]. Approximative unbounded degree interpolation arises also naturally in the issues of computational learnability of sparse rational functions (cf. [KW 89]). Another version of unbounded degree univariate polynomial interpolation was studied in [BT 89].

To bound the exponents appearing in some t-sparse representation of a t-sparse rational function f(X) of one variable, we will proceed as follows. We consider representations of f(X) of the form $(\sum_{i=1}^{t} a_i X^{\alpha_i})/(\sum_{i=1}^{t} b_i X^{\beta_i})$, where the a_i and b_i are real numbers and the α_i and β_i are non-negative real numbers. Such a function is called a *quasirational* function. We show that for t-sparse f(X) the α_i and β_i must satisfy a system S of polynomial equalities and inequalities whose coefficients depend on the value of f(X)at $2(t+1)^t - 1$ points. By evaluating the black box for f(X) at these points, we can determine this system. Using the results of [GV 88], we can bound a *real* solution of this system. Using the fact that f(X) is a t-sparse *rational* function, we are then able to bound an integer solution of S and this gives our desired bound.

The rest of this paper is organized as follows.

In Section 1 we give a formal definition of quasirational functions and prove some basic facts about these functions. In Section 2, we describe some elementary properties of left euclidean rings. An example of such a ring is $F[\mathcal{D}]$, where F is the field of quasirational functions of one variable and \mathcal{D} is the operator defined by $\mathcal{D}(f(X)) = f(pX)$ for some fixed prime p. For this ring, we are able to derive an analogue of the Sylvester matrix and the resultant. In Section 3 we use this to obtain the system S and the bound for the exponents appearing in a t-sparse representation of a t-sparse rational function. In Section 4 we show how the results of section 3 can be used to obtain a bound on the exponents of a t-sparse rational function of several variables. In Section 5 we describe an algorithm to interpolate t-sparse rational functions and give complexity bounds.

1. Quasirational Functions.

A finite sum of the form

 $\sum_{I} c_{I} \mathbf{X}^{I}$

where $I = (\alpha_1, \ldots, \alpha_n)$, $0 \le \alpha_i \in \mathbb{R}$, $\mathbf{X}^I = X_1^{\alpha_1} \cdot \ldots \cdot X_n^{\alpha_n}$, $c_I \in \mathbb{R}$ is called a *quasipolynomial* of *n* variables. Denote by $\mathbb{R} < X_1, \ldots, X_n >$ the ring of quasipolynomials of *n* variables.

A ratio of two quasipolynomials is called a *quasirational* function. If the number of terms in the sum is at most t, we say that the quasipolynomial is t-sparse. If a quasirational function can be represented as a ratio of two t-sparse quasipolynomials, we say that it is also t-sparse. We use the expressions "polynomial" or "rational function" in this usual sense, that is for quasipolynomials or quasirational functions with non-negative integer exponents in their terms.

We assume that we are given an *n*-variable rational function f with integer coefficients into which we can put points with rational coefficients. The output of the black box is either the value of the function at this point or some special sign, e.g. " ∞ ", if the denominator of the irreducible representation of the function vanishes at this point (a representation f = g/h, $g, h \in I\!\!R[X_1, \ldots, X_n]$, is irreducible if g and h are relatively prime). In what follows, we will sometimes obtain in intermediate steps a representation of a rational function in the form of a quasirational function. Nevertheless, our aim is to obtain a representation of a rational function in the usual form, provided that it is t-sparse.

We will need a zero test for t-sparse rational functions. This is similar to well known

zero tests for t-sparse polynomials (cf. [GK 87], [GKS 88], [BT 88]). Recall that if M_1, \ldots, M_t are distinct positive numbers, then any $t \times t$ subdeterminant of the $(2t-1) \times t$ matrix $(M_{\bullet}^j)_{1 \leq s \leq t, 1 \leq j \leq 2t-1}$ is non-singular (c.f. [EI 76]).

To test if a t-sparse rational function f is identically zero, use its black box to evaluate f at the 2t-1 points $P^j = (p_1^j, \ldots, p_n^j), 1 \le j \le 2t-1$, where the p_1, \ldots, p_n are distinct primes. Since the black box gives output based on an irreducible representation of f, we see that any zero of the denominator of such a representation is a zero of the denominator of a t-sparse representation of f. Using the remark about the matrix (M_i^j) above we see that the denominator can vanish at, at most, t-1 of these points. The same concerns the numerator. Therefore, the t-sparse function f is not identically zero if and only if the black box outputs a number different from 0 and ∞ at one of the points P^j .

The next result concerns different t-sparse representations of a quasirational function f. This result can be thought of as saying that, under suitable hypotheses, two such representations can only differ in certain redundant terms that can be eliminated. If g is a quasipolynomial, we denote by $\operatorname{ord}_{X_i}(g)$ the least power of X_i occurring in g. We call a representation $g_1/h_1 = f$ normalized if for each $i, 1 \leq i \leq n$, min $(\operatorname{ord}_{X_i}(g_1), \operatorname{ord}_{X_i}(h_1)) = 0$. For an arbitrary g_1/h_1 , there is a unique monomial M such that $(g_1/M)/(h_1/M)$ is normalized. We call the latter representation the normalization of g_1/h_1 .

Lemma 1. Assume that g_1/h_1 is a t-sparse representation of a quasirational function and $g_2/h_2 = g_1/h_1$ is another t-sparse normalized representation. Let $d = \max_i \{ \deg_{X_i}(g_1), \deg_{X_i}(h_1) \}$. We can delete some terms from g_2 and h_2 obtaining \bar{g}_2, \bar{h}_2 so that

$$\bar{g_2} / \bar{h_2} = g_1 / h_1$$

and

$$\max_{i} \{ \deg_{X_i}(\bar{\bar{g}}_2), \deg_{X_i}(\bar{\bar{h}}_2) \} \le 2td,$$

where $\bar{\bar{g}}_2 / \bar{\bar{h}}_2$ is the normalization of $\bar{g}_2 / \bar{\bar{h}}_2$.

PROOF.

Write

$$g_2 = \sum_i g_2^{(i)} X_1^{\beta_i}, \quad h_2 = \sum_i h_2^{(j)} X_1^{\gamma_j}$$

where $\beta_1 < \beta_2 < \ldots, \gamma_1 < \gamma_2 < \ldots$ and $0 \neq g_2^{(i)} \in \mathbb{R} < X_2, \ldots, X_n >, 0 \neq h_2^{(j)} \in \mathbb{R} < X_2, \ldots, X_n >$. We can assume that $\beta_1 = 0$ (the argument is similar if $\gamma = 0$). In this case we see from $g_2h_1 = h_2g_1$, that $\gamma_1 \leq d$. If $\beta_{i+1} - \beta_i \leq d$ for each i, then $\deg_{X_1} g_2 \leq (t-1)d$. This would imply $\deg_{X_1} h_2 \leq \deg_{X_1} g_2h_1 \leq td$ and we would be done.

If $\gamma_{i+1} - \gamma_i \leq d$ for each *i*, then $\deg_{X_1} h_2 \leq td$. This would imply that $\deg_{X_1} g_2 \leq \deg_{X_1} h_2 g_1 \leq td + d \leq 2td$ and we would be done. Therefore we can assume that there is a minimal number *s* such that $\beta_{i_0} \leq s - d < s < \beta_{i_0+1}$ and $\gamma_{j_0} \leq s - d < s < \gamma_{j_0+1}$ for suitable i_0, j_0 . Since $\beta = 0$ and $\gamma_1 \leq d$, we have that $s \leq (t+1)d \leq 2td$.

Let

$$ilde{g}_2 = \sum_{i \leq i_0} g_2^{(i)} X_1^{eta_i}, \quad ilde{h}_2 = \sum_{j \leq j_0} h_2^{(j)} X_1^{\gamma_j}.$$

If one compares the coefficients of X_1^{ρ} , $\rho \leq s$, in $g_2h_1 = h_2g_1$, one can see that $\tilde{g}_2h_1 = \tilde{h}_2g_1$ so $\tilde{g}_2/\tilde{h}_2 = g_1/h_1$.

We now take the normalization $\tilde{\tilde{g}}_2/\tilde{\tilde{h}}_2$ of \tilde{g}_2/\tilde{h}_2 and apply considerations similar to those above to $\tilde{\tilde{g}}_2/\tilde{\tilde{h}}_2$ with X_2 playing the role of X_1 . At the end of this process we obtain the normalized representation $\bar{\tilde{g}}_2/\tilde{\tilde{h}}_2$. It corresponds to a pre-normalized \bar{g}_2/\tilde{h}_2 that satisfies the conclusion of the lemma.

Corollary. If, in the above Lemma, we assume $g_2, h_2 \in I\!\!R[X_1, \ldots, X_n]$ are polynomial, then we can conclude that $\bar{g}_2, \bar{h}_2 \in I\!\!R[X_1, \ldots, X_n]$ as well.

We note that in Lemma 1 and its corollary, \bar{g}_2 and \bar{h}_2 are obtained by eliminating terms of sufficiently high degree and keeping lower order terms in g_2 and h_2 .

2. Right Euclidean Rings (a digest).

Let F be a field and let $\mathcal{D}: F^+ \to F^+$ be a homomorphism with respect to the additive structure of F. Let $F[\mathcal{D}]$ be the subring of HOM (F^+, F^+) generated by F (acting on F^+ by multiplication) and \mathcal{D} . We assume that each element $a \neq 0$ from $F[\mathcal{D}]$ can be uniquely represented in the form $a = \sum_{0 \le i \le m} \alpha_i \mathcal{D}^i$ where $\alpha_i \in F$ and $\alpha_m \ne 0$. We denote the integer m by deg(a) and adopt the convention that deg(0) = $-\infty$.

We furthermore assume that for $a, b \in F[\mathcal{D}]$, $\deg(ab) = \deg(a) + \deg(b)$. This assumption is equivalent to the statement that for each α in F there are unique α_1, α_2 in F, with $\alpha_1 \neq 0$, such that $\mathcal{D} \cdot \alpha = \alpha_1 \mathcal{D} + \alpha_2$. We can conclude that there exists right Euclidean division in $F[\mathcal{D}]$, that is, for any $a, b \in F[\mathcal{D}]$ $b \neq 0$, there exist unique b_1, b_2 with $\deg(b_2) < \deg(b)$ such that $a = b_1b + b_2$. This leads to a right Euclidean algorithm and a notion of greatest common right divisor $(\gcd(a, b))$ of two elements a and b, which can be represented in the form $\gcd(a, b) = a_1a + b_1b$ for some $a_1, b_1 \in F[\mathcal{D}]$. Furthermore $a = a_0 \gcd(a, b)$ and $b = b_0 \gcd(a, b)$ for some $a_0, b_0 \in F[\mathcal{D}]$.

Let deg(a) = m and deg(b) = k and consider

$$\mathcal{D}^i(a) = \sum_{0 \le j \le m+i} a^{(i)}_j \mathcal{D}^j, \quad \mathcal{D}^l(b) = \sum_{0 \le j \le k+l} b^{(l)}_j \mathcal{D}^j$$

for $0 \le i \le k - 1, 0 \le l \le m - 1$.

Let S be the $(m + k) \times (m + k)$ matrix whose columns correspond to the operators $\mathcal{D}^{k+m-1}, \ldots, \mathcal{D}^2, \mathcal{D}, 1$ and whose rows contain the coefficients of the operators in $\mathcal{D}^i(a)$, $0 \leq i \leq k-1$ and $\mathcal{D}^i(b)$, $0 \leq l \leq m-1$ (S resembles the Sylvester matrix [VDW 66]; for differential operators a similar object is described in [G 88]). As in [G 88], one can show:

Lemma 2. $\deg(\operatorname{gdrc}(a, b)) = n - \operatorname{rank}(S).$

In what follows we restrict ourselves to the case where F is the field of quasirational functions in one variable and \mathcal{D} is the operator defined by $\mathcal{D}(X^{\alpha}) = (pX)^{\alpha}$, where p is some fixed prime number. Note that $\mathcal{D} \cdot f = \mathcal{D}(f) \cdot \mathcal{D}$.

Lemma 3. If $f \in F$ and $\mathcal{D}(f) = f$, then $f \in \mathbb{R}$.

PROOF. If $\mathcal{D}(f) = f$, then $f(X) = f(pX) = f(p^2X) = \dots$ The zero test of section 1 implies that f(X) = f(YX) for a new variable Y. If f = g/h let

$$g = \sum a_i X^{\alpha_i}, h + \sum b_i X^{\beta_i}$$

 $0 \leq \alpha_1 < \alpha_2 < \dots$ $0 \leq \beta_1 < \beta_2 < \dots$, and $a_i, b_i \in I\!\!R$. Since

$$g(YX)h(X) = g(X)h(YX),$$

we can conclude, by comparing coefficients of the corresponding monomials in X and Y, that $\alpha_1 = \beta_1, \alpha_2 = \beta_2, \ldots$ and $a_i b_j = a_j b_i$ for all i, j. Therefore $f \in \mathbb{R}$. \Box

Lemma 4. If $y_1, \ldots, y_n \in F$, then y_1, \ldots, y_n are linearly dependent over $I\!R$ if and only if

$$W(y_1,...,y_n) = \det \left[egin{array}{cccc} y_1(x) & \ldots & y_n(x) \ y_1(px) & \ldots & y_n(px) \ dots & dots \ y_1(p^{n-1}x) & \ldots & y_n(p^{n-1}x) \end{array}
ight] = 0$$

PROOF. If y_1, \ldots, y_n are linearly dependent over $I\!R$ then, clearly, $W(y_1, \ldots, y_n) = 0$. Now assume that $W(y_1, \ldots, y_n) = 0$. In this case there exist $f_1, \ldots, f_n \in F$, not all zero, such that

$$f_1y_1+\ldots+f_ny_n=f_1\mathcal{D}y_1+\ldots+f_n\mathcal{D}y_n=f_1\mathcal{D}^{n-1}y_1+\ldots+f_n\mathcal{D}^{n-1}y_n.$$

We may assume $f_1 = 1$. Applying \mathcal{D} to each of these equations, we have

$$\mathcal{D}^i y_1 + \mathcal{D} f_2 \mathcal{D}^i y_1 + \ldots + \mathcal{D} f_n \mathcal{D}^i y_n = 0$$

for $i = 1, \ldots, n$. This implies that

$$(f_2 - \mathcal{D}f_2)\mathcal{D}^i y_2 + \ldots + (f_n - \mathcal{D}f_n)\mathcal{D}^i y_n = 0$$

for i = 1, ..., n - 1. Either $f_i - \mathcal{D}f_i = 0$ for i = 2, ..., n, in which case we are done by Lemma 3, or by induction these exist $\alpha_2, ..., \alpha_n \in \mathbb{R}$, not all zero, such that $\alpha_2 \mathcal{D}y_2 + ... + \alpha_n \mathcal{D}y_n = 0$. Therefore $\mathcal{D}(\alpha_2 y_2 + ... + \alpha y_n) = 0$ so $\alpha_2 y_2 + ... + \alpha_n y_n = 0$. \Box

Corollary. Let $L = \sum_{i=0}^{t} a_i \mathcal{D}^i$ with $a_i \in F$, not all zero. The dimension of the *IR*-vectorspace of solutions in F of Ly = 0 is at most t.

PROOF. Let y_1, \ldots, y_{t+1} be solutions of Ly = 0. We then have

$$(a_0,\ldots,a_t) \left[egin{array}{cccc} y_1&\ldots&y_{t+1}\\ \mathcal{D}y_1&\ldots&\mathcal{D}y_{t+1}\\ dots&dots\\ \mathcal{D}^ty_1&\ldots&\mathcal{D}^ty_{t+1} \end{array}
ight] = 0$$

Lemma 4 implies that y_1, \ldots, y_{t+1} are linearly dependent.

Lemma 5. Let $L = \sum_{j=0}^{t} a_j \mathcal{D}^j$ with $a_i \in \mathbb{R}$ and assume that $P_L(z) = a_t z^t + \ldots + a_0 \in \mathbb{R}[z]$ has t distinct roots ≥ 1 , say $p^{\alpha_1}, \ldots, p^{\alpha_t}$. Then $\{X^{\alpha_1}, \ldots, X^{\alpha_n}\}$ is a base for the space of solutions of Ly = 0.

PROOF. One easily sees that $L(X^{\alpha_i}) = 0$ for i = 1, ..., t. The functions $X^{\alpha_1}, ..., X^{\alpha_t}$ are linearly independent over $I\!R$, so by the corollary to Lemma 4 they must be a basis of the space of solutions.

Lemma 6. Let L be as above and assume that $L = L_1 \cdot L_2$ where $L_1 = \sum_{j=0}^{t-s} b_j^{(1)} \mathcal{D}^j$ and $L_2 = \sum_{j=0}^{s} b_j^{(2)} \mathcal{D}^j$ with $b_j^{(i)} \in F$. Then the space of solutions in F of $L_2(y) = 0$ has dimension s.

PROOF. Let V be the solution space of Ly = 0. By Lemma 5, this has dimension t. L_2 maps V into the solution space of L_1 , which has dimension at most t - s by Lemma 4. Therefore the dimension of the solution space of $L_2y = 0$ is at least s and so by Lemma 4, it must equal s.

3. Bounding the Exponents of a Sparse Univariate Rational Function

Lemma 5 in the previous section allows us to characterize t-sparse quasipolynomials gas those quasipolynomials for which there exits an operator of degree t, $L = \sum_{j=0}^{t} a_j \mathcal{D}^j$, with $P_L(z) = a_t z^t + \ldots + a_0 \in I\!\!R[z]$ having distinct real roots ≥ 1 , such that Lg = 0. Therefore a t-sparse quasirational function f is a quasirational function for which there exists a quasipolynomial h and operators of degree t, L_1 and L_2 as above such that $L_1(h) = 0$ and $L_2(hf) = 0$. $L_1(y)$ and $L_2(yf)$ will therefore have a common solution. The results of section 2 allow us to eliminate y using the determinant of the Sylvester matrix. This determinant is a quasirational function and, by evaluating at sufficiently many points, we obtain (together with the conditions that the α_i, β_j are distinct and ≥ 1) a system of polynomial inequalities that must be satisfied by the exponents appearing in f. We will then bound a *real* solution of this system using [GV 88] and, assuming that f is a univariate rational function, we can use Lemma 1 to bound the exponents of f.

We now proceed more formally. Let $f = \frac{g}{h}$ be a t-sparse quasirational function of one

variable where $g = \sum_{i=1}^{t} a_i X^{\alpha_i}$ and $h = \sum_{i=1}^{t} b_i X^{\beta_i}$ are t-sparse quasipolynomials. Let $G(z) = c_0 + c_1 z + \ldots + z^t$ be the unique monic polynomial whose roots are $p^{\alpha_1}, \ldots, p^{\alpha_t}$ and let $H(z) = d_0 + d_1 z + \ldots + z^t$ be the unique monic polynomial whose roots are $p^{\beta_1}, \ldots, p^{\beta_t}$. Consider the operators $L_G = \sum_{i=0}^{t} c_i \mathcal{D}^i$ and $L_H = \sum_{i=0}^{t} d_i \mathcal{D}^i$ (where $d_t = c_t = 1$). We then have $L_H(h) = 0$ and $L_G(fh) = 0$. Therefore $L_H(y) = 0$ and $L_G(fy) \equiv \tilde{L}_G(y) = 0$ have a non-zero common solution y = h in F (note that the coefficients of \tilde{L}_G are IR-linear combinations of $f, \mathcal{D}f, \ldots, \mathcal{D}^t f$). Consider the Sylvester matrix $S = S(c_0, c_1, \ldots, c_{t-1}, d_0, \ldots, d_{t-1}, f)$ of L_H and \tilde{L}_G . By Lemma 2, det (S) = 0 (note that det S is a quasirational function).

Conversely if det (S) = 0, then Lemma 2 implies that deg $(\operatorname{gcrd}(L_H, \tilde{L}_G)) \ge 1$. Since the coefficients of L_H satisfy the hypotheses of Lemma 5 and $\operatorname{gcrd}(L_H, \tilde{L}_G)$ divides L_H , L_H and \tilde{L}_G will have a common non-zero solution h_0 in F (by Lemma 6.). Lemma 5 then implies that f is a t-sparse quasirational function because h_0 and $h_0 f$ are both t-sparse quasipolynomials, again by Lemma 5. We have therefore proved the following lemma.

Lemma 7. A quasirational function f is t-sparse if and only if there exist real numbers $\bar{c}_0, \ldots, \bar{c}_{t-1}, \bar{d}_0, \ldots, \bar{d}_{t-1}$ such that

- (i) det $(S(\bar{c}_0,\ldots,\bar{c}_{t-1},\bar{d}_0,\ldots,\bar{d}_{t-1},f)) = 0$
- (ii) there exist t distinct real numbers ≥ 1 that are roots of

$$G(z) = \bar{c}_0 + \ldots + \bar{c}_{t-1}z^{t-1} + z^t = 0$$

and there exist t distinct real numbers ≥ 1 that are roots of

$$H(z) = \bar{d}_0 + \ldots + \bar{d}_{t-1}z^{t-1} + z^t = 0.$$

Now assume that f is a t-sparse rational function whose coefficients are integers. We see that each entry of S is a t-sparse rational function. From the form of the matrix, we see that det(S) is a $(t+1)^t$ sparse rational function. Therefore condition (i) is equivalent (by the zero test) to the fact that det $(S)_{X=p_1^t}$ is either ∞ of 0 for $i = 1, \ldots, 2(t + 1)^t - 1$ (p_1 is any prime). For at least $(t+1)^t$ of these points det $(S)_{X=p_1^t}$ will be zero. Using the black box, we can determine a system of $(t+1)^t$ equations in the unknowns $c_0, \ldots, c_{t-1}, d_0, \ldots, d_{t-1}$ of degree at most 2t, that is equivalent to the vanishing of det(S) at these points. Assume the bitsize of the values (yielded by the black box) of $\mathcal{D}^j f(p_1^i), i = 1, \ldots, 2(t+1)^t - 1, j = 0, \ldots, t$ is at most M (that is, those that are not ∞ and therefore rational numbers). We then see that the bitsize of each coefficient in this system is at most $O(t \ln t) + tM$.

Furthermore, condition (ii) is equivalent to

(ii') On $1 \le z \le 1 - c_{t-1} + c_{t-2} - \ldots \pm c_0$, the polynomial G(z) = 0 has precisely t roots and a similar statement holds for H. In addition, the discriminants of G and Hare not zero.

The first sentence of (ii') can be expressed in terms of Sturm sequences. This yields a system of 2t polynomial inequalities and (by the Habicht subresultant theorem) each inequality has degree at most 2t and the bitsize of the coefficients is at most $O(t \ln t)$. Similar bounds hold for the discriminants.

Therefore, under the assumption that f is a t-sparse rational function with integer coefficients, we are able to construct a system of polynomial inequalities equivalent to (i) and (ii) and bound the bitsize of the coefficients of this system. The results of [GV 88] imply that this system has a solution in a ball of radius $\exp((Mt^{t^2})^{O(1)})$.

This gives a bound on some solution $\bar{c}_0, \ldots, \bar{c}_{t-1}, \bar{d}_0, \ldots, \bar{d}_{t-1}$ which gives a t-sparse quasirational representation of f. The exponents in this representation can be bounded from these $\bar{c}_0, \ldots, \bar{c}_{t-1}, \bar{d}_0, \ldots, \bar{d}_{t-1}$ also by $\exp((Mt^{t^2})^{O(1)})$ since they are roots of the polynomials G(z) and H(z). By Lemma 2, the exponents in a t-sparse rational function representation of f are bounded by a similar number. This bound therefore can be explicitly calculated by making $2(t+1)^t - 1$ black box evaluations.

4. Bounding the Exponents of a Sparse Multivariate Rational Function.

Let

$$f(X_1,...,X_n) = \frac{g(X_1,...,X_n)}{h(X_1,...,X_n)} = \frac{\sum_{i=1}^t g_i(X_2,...,X_n) X_1^{\alpha_i}}{\sum_{i=1}^t h_i(X_2,...,X_n) X_1^{\beta_i}}$$

be a normalized representation of the t-sparse rational function f.

Consider the $2t^4 + 2t^2 + 1$ points $P^j = (p_2^j, ..., p_n^j)$ for $1 \le j \le 2t^4 + 2t^2 + 1$. Let

$$f^{j}(X_{1}) = f(X_{1}, p_{2}^{j}, \dots, p_{n}^{j}) = \frac{\sum_{i=1}^{t} g_{i}(p_{2}^{j}, \dots, p_{n}^{j}) X_{1}^{\alpha_{i}}}{\sum_{i=1}^{t} h_{i}(p_{2}^{j}, \dots, p_{n}^{j}) X_{1}^{\beta_{i}}}$$

Note that there are at most $2t^2$ points P^j for which some g_i or h_i vanishes at P^j . We call these points bad points. For a point P_j that is not bad, let D_j be the bound on the degree of some normalized t-sparse representation of $f^j(X_1)$. For these points, Lemma 2 allows is to conclude that there exist t_1, t_2 (not necessarily unique) such that

$$f^{j}(X_{1}) = \frac{\sum_{i=1}^{t_{1}} g_{i}(p_{2}^{j}, \dots, p_{n}^{j}) X_{1}^{\alpha_{i}}}{\sum_{i=1}^{t_{2}} h_{i}(p_{2}^{j}, \dots, p_{n}^{j}) X_{1}^{\beta_{i}}} = \frac{\bar{g}^{j}}{\bar{h}^{j}}$$

 \mathbf{and}

$$\max\{\deg_{X_1}\bar{g}^j, \deg_{X_1}\bar{h}^j\} \le 2tD_j.$$

To each j corresponds some pair t_1, t_2 . Therefore, at least $2t^2 + 1$ non-bad points P^j correspond to some pair $(\tilde{t}_1, \tilde{t}_2)$. For this pair $(\tilde{t}_1, \tilde{t}_2)$ we have that

$$g(X_2,\ldots,X_n)\sum_{i=1}^t h_i(X_2,\ldots,X_n)X_1^{\alpha^i} - h(X_1,\ldots,X_n)\sum_{i=1}^t g_i(X_2,\ldots,X_n)X_1^{\beta^i}$$
(1)

is zero at these $2t^2 + 1$ points. If we consider (1) as a polynomial in X_1 whose coefficients are $2t^2$ sparse polynomials in X_1, \ldots, X_n , we see that (1) is identically zero. This implies that f has a t-sparse representation with $\deg_{X_1} f \leq B_1 = \max_j \{2tD_j\}$. We consider this representation and let X_2 play the role of the principal variable. We apply the same construction to prove the existence of a representation with $\deg_{X_1} f \leq B_1$ and $\deg_{X_2} f \leq B_2$. In this way we are able to determine B for which these exists a t-sparse representation of f with $\deg_{X_1} f \leq B$ for $i \leq i \leq n$.

Note that as in the univariate case, $B \leq \exp((Mt^{t^2})^{O(1)})$ where M is a bound on the bitsize of $f(p_1^j, \ldots, p_n^j)$ for $1 \leq j \leq 2(t+1)^t - 1$.

5. Interpolation of Sparse Multivariate Rational Functions.

Let f be a t-sparse multivariate rational function and let $B \leq \exp((Mt^{t^2})^{O(1)})$ be the bound obtained in the previous section. Let

$$\mathcal{A} = \{A_i = (lpha_{i_1}, \dots, lpha_{i_n}) | 0 \le lpha_{ij} \le B\}$$

$$\mathcal{B} = \{B_i = (\beta_{i_1}, \ldots, \beta_{i_n}) | 0 \leq \beta_{ij} \leq B\}.$$

Select, in parallel, 2 t-tuples $I = \{A_1, \ldots, A_t\}$, $J = \{B_1, \ldots, B_t\}$ with $A_i \in \mathcal{A}$ and $B_i \in \mathcal{B}$. We calculate f(X) at (p_1^i, \ldots, p_n^i) for $i = 1, \ldots, 4t^2$. For each selection of I and J we obtain the following linear system

$$f(p_1^i,\ldots,p_n^i)(b_1(p_1^{\beta_{11}}\ldots p_n^{\beta_{1n}})^i+b_1(p_1^{\beta_{t1}}\ldots p_n^{\beta_{tn}})^i)=a_1(p_1^{\alpha_{11}}\ldots p_n^{\alpha_{1n}})^i+\ldots+a_t(p_1^{\alpha_{t1}}\ldots p_n^{\alpha_{tn}})^i$$
(2)

where $1 \le i \le 4t^2$, in the unknowns $b_1, \ldots, b_t, a_1, \ldots, a_t$.

If such a system has a solution $\bar{b}_1, \ldots, \bar{b}_t, \bar{a}_1, \ldots, \bar{a}_t$, then the zero test implies that

$$f(X_1,...,X_n) = \frac{\bar{a}_1 X_1^{\alpha_{11}} \dots X_n^{\alpha_{1n}} + \dots + \bar{a}_t X_1^{\alpha_{t1}} \dots X_n^{\alpha_{tn}}}{\bar{b}_1 X_1^{\beta_{11}} \dots X_n^{\alpha_{1n}} + \dots + \bar{b}_t X_1^{\alpha_{t1}} \dots X_n^{\alpha_{tn}}}$$

For some I and J we will be able to solve (2) so the algorithm terminates with a correct answer. We now give an analysis of the complexity. Each of A and B contain B^n terms. We select t elements from each, so there are $O(B^{nt})$ systems of type (2), each of size at most $4t^2$. This implies that the sequential time complexity is $B^{O(nt)}$ and the parallel complexity is $(nt \log B)^{O(1)}$ (cf. [BGH 82], [M 86], [KR 88]). We can further bound B in terms of the size of the output. Let $\delta = \max_i \{\deg_{X_i} f\}$ and let μ be a bound on the bitsize of the coefficients of f. Let μ_j be the bitsize of any coefficient of $f^j(X_1)$ (as in section 4). We then have that the bitsize of any output is not greater than $\mu_j + t + \delta$. Furthermore, each μ_j can be bounded by $\mu + O(t^4 \delta n \log n)$ by looking at each term of the representation of f and noting that $p_n = O(n \log n)$. Therefore

$$B_{1} \leq \exp((Mt^{t^{2}})^{O(1)})$$

$$\leq \exp(((\mu + t^{4}\delta n \log n + \delta)t^{t^{2}})^{O(1)})$$

$$= \exp(((\mu + \delta n \log n)t^{t^{2}}))^{O(1)})$$

Therefore the sequential complexity of the algorithm is $\exp((\mu + \delta n \log n)t^{t^2})^{O(1)}$ and the parallel complexity is $((\mu + \delta n \log n)t^{t^2})^{O(1)}$. Therefore, for fixed t, interpolation of t-sparse rational functions can be done in polynomial parallel time (and sequential storage).

and

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