

# Using an Efficient Sparse Minor Expansion Algorithm to Compute Polynomial Subresultants and the Greatest Common Denominator

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**Abstract**—In this paper, the use of an efficient sparse minor expansion method to directly compute the subresultants needed for the greatest common denominator (GCD) of two polynomials is described. The sparse minor expansion method (applied either to Sylvester's or Bezout's matrix) naturally computes the coefficients of the subresultants in the order corresponding to a polynomial remainder sequence (PRS), avoiding wasteful recomputation as much as possible. It is suggested that this is an efficient method to compute the resultant and GCD of sparse polynomials.

**Index Terms**—Inners, minor expansion, polynomial GCD, subresultants, sparse matrices, sparse polynomials.

## I. INTRODUCTION

THE EMPIRICAL study of polynomial resultants by Ku and Adler [1], the discussions of the subresultant greatest common denominator (GCD) algorithm by Brown [2], [3], and the use of "inners" (subresultants) [4] for a multivariable polynomial greatest common factor (GCF) by Bose [5] suggest that the direct application of minor expansion to either the Sylvester or Bezout matrices is an efficient and interesting method of computing the GCD of two sparse polynomials.

It will be shown in this paper that the process of computing the subminors required in an efficient sparse minor algorithm [6], [7] quite easily and naturally yields the requisite subresultants and GCD. The crux lies in an appropriate reordering of either Sylvester's or Bezout's matrix, very similar to that studied in [7].

A number of authors have studied the problems of computing determinants of matrices with polynomial elements. When the polynomials are sparse (many zero coefficients in a full expansion), minor expansion is significantly more efficient than elimination methods [8], [9]. A particularly effective iterative algorithm [8] processes the rows of an  $N \times N$  matrix in order, computing the minors of all of the  $m \times m$  submatrices consisting of the first  $m$  rows and  $m$  of the  $N$  columns. This set of  $m \times m$  minors is then combined (with due attention to sign) with the elements of the  $m + 1$ st row to produce a new set of  $(m + 1) \times (m + 1)$  minors.

When the matrix itself is sparse (many zero elements), a careful choice of the order in which the rows are processed can have a significant effect on the total storage and time

required to compute the determinant. An efficient sparse minor expansion algorithm, using a sparse matrix representation and careful row ordering has been discussed by the author [6], [7].

## II. SUBRESULTANTS AND SYLVESTER'S MATRIX

The method will be described using the notation of Brown and Traub [3], which we briefly review here. We are interested in computing the resultant or GCD of two multivariate polynomials,  $F$  and  $G$ . Expand  $F$  and  $G$  (assumed to be primitive) with respect to a "main variable,"  $x$ :

$$F(x) = \sum_{i=0}^{\psi} f_i x^{\psi-i}, \quad G(x) = \sum_{j=0}^{\gamma} g_j x^{\gamma-j}, \quad \text{with } \psi \geq \gamma. \quad (1)$$

The coefficients  $\{f_i, g_i\}$  are polynomials in other variables, and  $\psi = \partial(F)$ ,  $\gamma = \partial(G)$  are the degrees of the polynomials.

Brown and Traub define the subresultant,  $S_j(F, G)$  as

$$= \det \begin{bmatrix} f_0 & f_1 & \cdots & f_{\gamma-j-1} & \cdots & f_{\psi+\gamma-2j-2} & x^{\gamma-j-1}F \\ & f_0 & \cdots & f_{\gamma-j-2} & & f_{\psi+\gamma-2j-1} & x^{\gamma-j-2}F \\ & & & f_0 & & f_{\psi+\gamma-2j} & F \\ g_0 & g_1 & \cdots & g_{\psi-j-1} & & g_{\psi+\gamma-2j-2} & x^{\psi-j-1}G \\ & g_0 & \cdots & g_{\psi-j-2} & & g_{\psi+\gamma-2j-1} & x^{\psi-j-2}G \\ & & & g_0 & & g_{\psi-j-1} & G \end{bmatrix} \quad (2)$$

where  $f_k = 0$  if  $k > \psi$ ,  $g_k = 0$  if  $k > \gamma$ , and  $0 \leq j < \min(\psi, \gamma) = \gamma$ . (This matrix is the transpose of that given in [3].)

Expanding the polynomials in the last column, it is shown that

$$S_j(F, G) = \sum_{l=0}^j s_{jl} x^{j-l}, \quad \text{for } 0 \leq j < \min(\psi, \gamma) = \gamma. \quad (3)$$

The coefficients,  $s_{jl}$ , are determinants of the  $(\psi + \gamma - 2j) \times (\psi + \gamma - 2j)$  matrices,  $S_{jl}$ :

$$S_{jl} = \begin{bmatrix} f_0 & f_1 & f_2 & \cdots & f_{\psi+\gamma-2j-2} & f_{\psi+\gamma-2j+l-1} \\ & f_0 & f_1 & \cdots & f_{\psi+\gamma-2j-3} & f_{\psi+\gamma-2j+l-2} \\ & & \vdots & & \vdots & \vdots \\ & & & f_0 & \cdots & f_{\psi-j-1} & f_{\psi-j+l} \\ g_0 & g_1 & g_2 & \cdots & g_{\psi+\gamma-2j-2} & g_{\psi+\gamma-2j+l-1} \\ & g_0 & g_1 & \cdots & g_{\psi+\gamma-2j-3} & g_{\psi+\gamma-2j+l-2} \\ & & \vdots & & \vdots & \vdots \\ & & & g_0 & \cdots & g_{\psi-j-1} & g_{\psi-j+l} \end{bmatrix} \begin{matrix} \uparrow \\ (\gamma-j) \text{ rows} \\ \downarrow \\ (\psi-j) \text{ rows} \\ \downarrow \end{matrix} \quad (4)$$

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identify  $s_{ji}$  by the key

$$\begin{aligned} \text{KEY}(s_{ji}) &= \text{KEY}(0, 1, \dots, \psi + \gamma - 2j - 2, \\ &\quad l + \psi + \gamma - 2j - 1) \\ &= 2^0 + 2^1 + \dots + 2^{x-1} + 2^{x+l} \\ &= (2^x - 1) + 2^x \cdot 2^l, \\ &\quad l = 0 \dots j \text{ with } x = \psi + \gamma - 2j - 1. \end{aligned} \quad (9)$$

Recall that  $j = \gamma - k$ ,  $k = 1 \dots \gamma$  so that

$$\text{KEY}(s_{ji}) = (B_k - 1) + B_k \cdot 2^l$$

where

$$\begin{aligned} B_k &= 2^{\psi+\gamma+2k-2\gamma-1} \\ &= 2^{\psi-\gamma+2k-1} = 4 \cdot B_{k-1}. \end{aligned} \quad (10)$$

Now  $B_0 \equiv 2^{\psi-\gamma-1}$  and is computed while computing the "base" set:

```

B := 1/2;
SET := NIL;
FOR I = 1: (ψ - γ) DO
    <<SET := DMRG1 (SET, G);
    G := SHIFT (G, 2);
    B := 2 * B>>;
    
```

where  $\text{SHIFT}(G, 2)$  simply "shifts" all keys of  $G$  by multiplying by 2. Now the main loop begins:

$$J := \gamma - 1;$$

```

STEP: NSET := DMRG1 (DMRG1 (SET, F), G);
IF NSET = NIL THEN GOTO DONE;
SET := NSET; B := 4 * B;
IF J = 0 THEN GOTO DONE;
F := SHIFT (F, 2); G := SHIFT (G, 2);
J := J - 1;
GOTO STEP;
    
```

Finally we compute the coefficients  $s_{ji}$ :

```

DONE: B1 := B - 1;
GCD := NIL;
FOR L := 0: J DO
    <<GCD := LOOKUP (B1 + B, SET) · GCD;
    B := 2 * B>>;
GCD := REVERSE GCD;
    
```

which computes the GCD in a nonsparse form. The actual program is given in the Appendix, in `MODE REDUCE` [7], [10]. (`MODE REDUCE` is a computer algebra system with a Pascal-like type definition facility.)

In order to find the GCD of nonprimitive polynomials, the above algorithm will be applied recursively to the coefficients, to obtain  $\text{GCD}(\text{cont}(F_1), \text{cont}(F_2))$ , as suggested in [5], so that finally,

$$\begin{aligned} \text{GCD}(F_1, F_2) \\ = \text{GCD}(\text{pp}(F_1), \text{pp}(F_2)) \cdot \text{GCD}(\text{cont}(F_1), \text{cont}(F_2)) \end{aligned} \quad (11)$$

Once the "base" set has been computed, subsequent sets of

minors always involve a joint shift of  $F$  and  $G$ :

$$\begin{aligned} \text{NSET} &\leftarrow \text{DMRG1}(\text{DMRG1}(\text{SET}, F), G); \\ F &\leftarrow \text{SHIFT}(F, 2); G \leftarrow \text{SHIFT}(G, 2); \end{aligned} \quad (12)$$

If we first make up a set of  $2 \times 2$  minors,  $FG \leftarrow \text{DMRG1}(F, G)$ , and generalize  $\text{DMRG1}$  to accept 2 sets of minors (as suggested in [7]), we find a simpler form

$$\begin{aligned} \text{NSET} &\leftarrow \text{DMRG2}(\text{SET}, FG); \\ FG &\leftarrow \text{SHIFT}(FG, 2); \end{aligned} \quad (13)$$

where the  $2 \times 2$  minors in  $FG$  are therefore computed only once, saving repeated computation.

It is interesting to note that the elements of Bezout's matrix (a  $\psi \times \psi$  matrix derived from Sylvester's matrix) are simple sums of the  $2 \times 2$  minors in  $FG$ . This suggests that the above method be directly applied to Bezout's matrix, producing an even more efficient method, discussed in the next section.

#### IV. BEZOUT'S MATRIX

The following  $\psi \times \psi$  matrix  $B$  (known as Bezout's matrix) has been used by Ku and Adler [1] to compute the resultant,  $S(F, G) = \det(B)$ :

$$B = \begin{bmatrix} g_0 & g_1 & g_2 & \dots & g_\gamma \\ & g_0 & g_1 & \dots & g_\gamma \\ & & \ddots & & \\ & & & g_0 & g_1 & \dots & g_\gamma \\ A_{11} & A_{12} & & & A_{13} & \dots & A_{1\psi} \\ \vdots & & & & & & \\ A_{\gamma 1} & A_{\gamma 2} & & & & & A_{\gamma\psi} \end{bmatrix} \begin{matrix} \uparrow \psi - \gamma \text{ base rows} \\ \downarrow \gamma \text{ rows} \end{matrix} \quad (14)$$

The  $A_{ij}$  are sums of  $2 \times 2$  minors,  $D_{ij}$ ,

$$D_{ij} = f_i g_j - f_j g_i \quad (15)$$

(where  $f_i$  or  $g_i = 0$  if  $i$  is out of range); and

$$A_{ij} = \sum_{k=0}^w D_{k, i+k} \quad (16)$$

with

$$w = \min(i - j, \psi - i - 1, j - 1, \psi - j - 1).$$

The  $A_{ij}$  are obtained from the coefficients of  $x$  in a homogeneous system of equations [1], just as Sylvester's matrix is obtained from the system of equations

$$\begin{bmatrix} x^{\psi-1} G(x) = 0 \\ x^{\psi-2} G(x) = 0 \\ \vdots \\ x^0 G(x) = 0 \\ x^{\gamma-1} F(x) = 0 \\ \vdots \\ x^0 F(x) = 0 \end{bmatrix}. \quad (17)$$



Each of the determinants  $s_{jl}$  now corresponds to a determinant of  $M_j$ , with  $l$  selecting one of the last  $j$  columns. The excess triangular portion of  $G$  rows can be replaced by 1 (after dividing by the  $g_0^{j-1}$  factors) so that the determinants  $s_{jl}$  are obtained from the matrix

$$B_j = \begin{bmatrix} g_0 & g_1 & \cdots & g_j & 0 & \cdots & 0 \\ & g_0 & & & g_j & & \\ & & \ddots & & & \ddots & \\ & & & g_0 & & & g_j \\ D_{01} & & & \cdots & & & D_{0\psi} \\ D_{02} & (D_{03} + D_{12}) & & \cdots & & & D_{1\psi} \\ \vdots & & & & & & \vdots \\ D_{0\gamma} & \cdots & & & & & \end{bmatrix} \quad (25)$$

applying the stepwise method of Section III to Bezout's matrix  $B$ . A program can be easily devised to compute the  $A_{ij}$  elements as the minors are generated in a stepwise fashion.

#### IV. CONCLUSIONS

The application of the efficient sparse minor expansion algorithm to the direct computation of the GCD via generation of all requisite subresultants is quite simple and efficient.

I am grateful for many enjoyable discussions on this subject with D. Yun, M. Rothstein, and S. Brown. The method can be applied either to Sylvester's matrix or to Bezout's reduced matrix, leading to an even more efficient method. When the input polynomials are sparse, Sylvester's matrix and even Bezout's matrix will be rather sparse, clearly motivating the use of a *sparse* minor expansion method.

#### APPENDIX USE OF THE EFFICIENT SPARSE MINOR ALGORITHM TO COMPUTE GCD

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Load SPARSE MINOR package (ref. 7) to define sparse minor data structures and procedures:

```
MODE ELEM =POLYNOMIAL, EROW=LISTOF ELEM;      Dense row
MODE DELEM=STRUCT(KEY:INTEGER,DETM:ELEM),      Sparse row
      DROW =LISTOF DELEM;
```

```
Procedures:  DMRG1(drow,drow)->drow          minor expansion
              SHIFT(drow,integer)->drow       Shift columns
```

---

```
ELEM PROCEDURE LOOKUP(K,D);
% Lookup K'th entry in minor list D;
DCL K:INTEGER, D:DROW;
IF NULL D THEN 0
ELSE IF KEY(HEAD D)=K THEN DETM(HEAD D)
ELSE LOOKUP(K,TAIL D);

EROW PROCEDURE EGCD(E1,E2);
% Generate minors in "good-order" to obtain Subresultants or GCD;
BEGIN
  DCL PHI,GAMMA,M,B1,B,J,L:INTEGER, E,GCD:EROW,
  F,G,SET,NSET:DROW;
  % Ensure PHI>=GAMMA;
  PHI:=LENGTH E1 -1; GAMMA:=LENGTH E2-1;
  IF GAMMA > PHI THEN <<M:=PHI; PHI:=GAMMA; GAMMA:=M;
  E:=E1; E1:=E2; E2:=E>>;
  % Convert to sparse rows;
  G:=DROW E1; F:=DROW E2;
  M:=PHI-GAMMA;
  % Block of PHI-GAMMA "short" rows;
  IF M>0 THEN <<SET:=F; M:=M-1; E:=1;
  FOR I:=1:M DO <<F:=SHIFT(F,2);
  B:=2*B;
  SET:=DMRG1(SET,F)>>;
  F:=SHIFT(F,2) >>
  ELSE <<SET:=DMRG1(F,G); GAMMA:=GAMMA-1; E:=2;
  G:=SHIFT(G,2); F:=SHIFT(F,2)>>;
  J:=GAMMA;
STEP:  % Merge successive rows, until NSET is empty;
  NSET:=DMRG1(DMRG1(SET,F),G);
  IF NULL NSET THEN GOTO DONE;
  SET:=NSET;
  IF J=0 THEN GOTO DONE;
  % Shift columns by 1 place;
  G:=SHIFT(G,2); F:=SHIFT(F,2);
  B:=4*B; J:=J-1;
  GOTO STEP;
DONE:  % Produce list of subresultants (or GCD);
  B1:=E-1;
  GCD:=NIL;
  FOR L:=0:J DO
    <<GCD:=LOOKUP(B1+B,SET) . GCD;
    B:=2*B>>;
  RETURN REVERSE GCD
ENE;
```

## REFERENCES

- [1] S. Y. Ku and R. J. Adler, "Computing polynomial resultants: Bezout's determinant vs. Collins' reduced PRS algorithm," *Commun. Ass. Comput. Mach.*, vol. 12, p. 23, 1969.
- [2] W. S. Brown, "On the subresultant PRS algorithm," presented at SYMSAC 76, Yorktown Heights, 1976. Bell Lab. Comput. Sci. Tech. Rep. 56, 1976. (Submitted to *Ass. Comput. Mach. TOMS*.)
- [3] W. S. Brown and J. F. Traub, "On Euclid's algorithm and the theory of subresultants," *J. Ass. Comput. Mach.*, vol. 18, p. 505, 1971.
- [4] E. I. Jury, "The theory and applications of the inners," *Proc. IEEE*, vol. 63, pp. 1044-1068, 1975.
- [5] N. K. Bose, "An algorithm for GCF extraction from two multi-variable polynomials," *Proc. IEEE (Letters)*, pp. 185-186, Jan. 1976.
- [6] M. L. Griss, "The algebraic solution of sparse linear systems via minor expansion," *Ass. Comput. Mach. TOMS*, vol. 2, p. 39, 1976.
- [7] —, "An efficient sparse minor expansion algorithm," *Proc. Ass. Comput. Mach. 76*, p. 429, 1976.
- [8] W. M. Gentleman and S. C. Johnson, "Analysis of algorithms, a case study: Determinants of polynomials," *Ass. Comput. Mach. TOMS*, vol. 2, p. 232, 1976.
- [9] E. Horowitz and S. Sahni, "On computing the exact determinant of matrices with polynomial entries," *J. Ass. Comput. Mach.*, vol. 22, p. 38, 1975.
- [10] A. C. Hearn, "A mode analyzing algebraic manipulation program," *Proc. Ass. Comput. Mach. 74*, p. 722, 1974.
- [11] M. L. Griss, "The definition and use of data-structures in REDUCE," *Proc. SYMSAC 76, Ass. Comput. Mach.*, p. 53, 1976.



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