Parallel computations of Gröbner bases in the Weyl algebra Something to run on a machine with 128 cores

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Institute for Mathematics and its Applications, Minneapolis

MSRI, Berkeley, 2007

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What is Weyl algebra?

Definition (n-th Weyl algebra over field K of characteristic 0)

$$D = A_n(K) = K\langle x, \partial \rangle = K\langle x_1, \partial_1, \dots, x_n, \partial_n \rangle,$$

where $[\partial_i, x_i] = \partial_i x_i - x_i \partial_i = 1$ and all other pairs commute.

Multiplication in Weyl algebra: Leibnitz rule

 $A_n = K\langle x_1, \ldots, x_n, \partial_1, \ldots, \partial_n \rangle$ then for $P, Q \in A_n$

$$PQ = \sum_{\alpha \in \mathbb{Z}_{\geq 0}^n} \frac{1}{\alpha!} \operatorname{Diff}(P, \partial^{\alpha}) * \operatorname{Diff}(Q, x^{\alpha}),$$

where Diff is a formal partial derivative (as if P, Q are polynomials) and * is the polynomial multiplication.

Weyl algebra in computer algebra systems

kan/sm1, risa/asir (Takayama, Noro); Macaulay 2 (Grayson, Stillman), *D*-modules for M2 (A.L., Tsai); Singular/Plural (Levandovskyy); CoCoA (group in Genova, Italy).

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Buchberger algorithm Parallel Buchberger

Let *R* be a Gröbner-friendly algebra (think: $R = K[x_1, \ldots, x_n]$).

Definition

Given a fixed *admissible* monomial ordering, a polynomial $f \in R$ has

- initial monomial lm(f);
- initial coefficient lc(f);
- initial term lt(f) = lc(f) lm(f).

Algorithm REDUCE(f, B)

In: $f \in R$, $B \subset R$ Out: a reduction of f w.r.t B

 $\begin{array}{l} f':=f\\ \text{WHILE } \exists g\in B \text{ such that } \mathrm{lm}(f') \text{ is divisible by } \mathrm{lm}(g) \text{; DO}\\ f':=f'-\frac{\mathrm{lt}(f')}{\mathrm{lt}(g)}\cdot g\\ \text{RETURN } f' \end{array}$

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Let $L(f,g) = \operatorname{lcm}(\operatorname{lm}(f),\operatorname{lm}(g))$.

Definition (s-polynomial of f and g)

$$sPoly(f,g) = \operatorname{lc}(g)\frac{L(f,g)}{\operatorname{Im}(f)}f - \operatorname{lc}(f)\frac{L(f,g)}{\operatorname{Im}(g)}g.$$

Definition

A set $G \subset R$ is a Gröbner basis of a left ideal $I \subset R$ if $I = R \cdot G$ and

 $\operatorname{gr}(R) \cdot \{LM(f) | f \in I\} = \operatorname{gr}(R) \cdot \{LM(g) | g \in G\},\$

where gr(R) is the graded ring associated to *R*.

Buchberger criterion

A set $G \subset R$ is a Gröbner basis if REDUCE(sPoly(f,g),G) = 0 for all $f,g \in G$.

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 G and the queue of s-pairs S;
- distributes orders to Slaves;
- collects results and updates G and S.

- stores a local basis G;
- receives orders from Master and send back the results;



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Key point: The order of *s*-pairs the same as in the serial algorithm. The strategies used for *s*-pair selection are preserved.

Implementation: C++ with MPI

- implemented from scratch in C++;
- uses MPI for communications;
- tested on clusters in the Minnesota Supercomputing Institute and NCSA.

Simulation of parallel computation

Assumptions:

- operations performed by Master are instantaneous;
- time for sending a package from one node to another depends linearly on its size.

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Faugére's F₄:

• can be adapted for Gröbner-friendly algebras;

- implementations: Noro (risa/asir) for Weyl algebra, Pearce Monagan (Maple) Ore algebras;
- loss of sparsity: multiplication of an operator by monomial increases the number of terms.

Example

Let $D = A_2 = K \langle x_1, x_2, \partial_1, \partial_2 \rangle$, $f_1 = x_1^2 \partial_2^3$, $f_2 = x_2^3 \partial_1^2$. F_4 starts computing a Gröbner basis of ideal $D \cdot \{f_1, f_2\}$ with the matrix

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... for Weyl algebra Loss of sparsity

The second step builds the following matrix:

	$x_1^2 x_2^3 \partial_1^2 \partial_2^3$	$x_1^2 x_2^2 \partial_1^2 \partial_2^2$	$x_1 x_2^3 \partial_1 \partial_2^3$	$x_1^2 x_2 \partial_1^2 \partial_2$	$x_2^3 \partial_2^3$	$x_2^3 \partial_1^2$	$x_1^2 \partial_2^3$	$x_1^2 \partial_1^2$
f_1	0	0	0	0	0	0	1	0
f_2	0	0	0	0	0	1	0	0
$f_1 f_2$	1	9	0	18	0	0	0	1
$f_2 f_1$	1	0	4	0	1	0	0	0

Help!

• Structured Gaussian elimination? ... (over a finite field) ...

• ... in parallel?

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s F₄ algorithm

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	$x_1^2 x_2^3 \partial_1^2 \partial_2^3$	$x_1^2 x_2^2 \partial_1^2 \partial_2^2$	$x_1 x_2^3 \partial_1 \partial_2^3$	$x_1^2 x_2 \partial_1^2 \partial_2$	$x_2^3 \partial_2^3$	$x_2^3 \partial_1^2$	$x_1^2 \partial_2^3$	$x_1^2 \partial_1^2$
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Gröbner basis Faugére's F₄ algorithm

Loss of sparsity

The second step builds the following matrix:

	$x_1^2 x_2^3 \partial_1^2 \partial_2^3$	$x_1^2 x_2^2 \partial_1^2 \partial_2^2$	$x_1 x_2^3 \partial_1 \partial_2^3$	$x_1^2 x_2 \partial_1^2 \partial_2$	$x_2^3 \partial_2^3$	$x_2^3 \partial_1^2$	$x_1^2 \partial_2^3$	$x_1^2 \partial_1^2$
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f_2	0	0	0	0	0	1	0	0
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- Parallel versions of Buchberger's algorithm can produce limited speedups;
- Our coarse-grain approach exhibits better speedups in the noncommutative algebra than in the (commutative) polynomial rings on "interesting" problems of similar size.
- It does make sense to use 128 nodes on this problem!

Future

- From theory to practice: a practically efficient parallel implementation is needed;
- Faugére's *F*₄ algorithm results in the loss of sparsity in the intermediate computation...
- ... however, it still feasible and its parallel version could be constructed;
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