

Parallel Algorithms in Algebraic Geometry

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Classical Parallelization Approaches

Parallel
Smoothness Test

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Parallel Algorithms

Smoothness

Smoothness II

Algorithm

Hybrid Approach

Summary

Modular Approaches

- ▶ Chinese remainder,

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- ▶ applications in algebraic geometry: e.g. Gröbner Bases,
normalization, integral bases

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inherently parallel structure need not be obvious

Recall: Jacobian criterion

$I = \langle f_1, \dots, f_s \rangle \subseteq \mathbb{C}[\underline{x}]$ equidimensional
 $X = V(I)$, $\dim(X) = n - c$

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in a smoothness test:

$$X \text{ non-singular} \iff \text{Sing}(X) = \emptyset$$

Jacobian criterion in parallel?

time consuming steps:

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Gröbner Basis, expensive,
very large input, small output

Hironaka's ν^* -Invariant

Defintion[Hironaka]

$(X, 0) \subset (\mathbb{A}_{\mathbb{C}}^n, 0)$ germ,

$I_{X,0} = \langle f_1, \dots, f_s \rangle \subset \mathbb{C}\{\underline{x}\}$

f_1, \dots, f_s SB of $I_{X,0}$, sorted by increasing order

$$\nu^*(X, 0) := (\text{ord}_0(f_1), \dots, \text{ord}_0(f_s))$$

Lemma[Hironaka]

$$(X, 0) \text{ non-singular} \iff \nu^*(X, 0) = \underbrace{(1, \dots, 1)}_{\text{codim}(X)}$$

Hironaka's criterion: computational viewpoint

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Comparison of the criteria:

Jacobian criterion computing singular locus

Hironaka's criterion testing non-singularity

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- ▶ locus of order at least 2 (if $= \emptyset$: non-singular)

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- ▶ locus of order at least 2 (if $= \emptyset$: non-singular)
- ▶ determine hypersurface of maximal contact (**local object**)
- ▶ descent in ambient dimension for considering next ideal

The Main Algorithm

Input:

- ▶ g polynomial
- ▶ $I_W = \langle g_1, \dots, g_r \rangle$ non-singular CI on $D(g)$
- ▶ $I_X = \langle f_1, \dots, f_s \rangle$, $I_W \subseteq I_X$

Output:

- ▶ True (X non-singular) or False (X singular)
1. if ($I_W == I_X$ on $D(g)$) return(True)
 2. if (not CheckOrder(I_W, I_X, g))return(False)
 3. list $L = \text{DescendOneStep}(I_W, I_X, g)$
 4. for ($I_U, I_{X|U}, g_U$) $\in L$ do
 - ▶ if (not SmoothnessTest($I_U, I_{X|U}, g_U$))
return(False)

The workhorses: CheckOrder

CheckOrder – only general idea

- ▶ Find regular system of parameters \underline{y} for $W \cap D(g)$ locally

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- ▶ compute

$$J = I_X + \left\langle \frac{\partial f_i}{\partial y_j} \right\rangle$$

(ideal of locus of order at least 2)

- ▶ test $V(J) == \emptyset$

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⇒ recombination step

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- ▶ covering by principal open subsets, splitting up the problem
- ▶ no recombination step possible

A Hybrid Approach

expensive parts of Hironaka style approach:

- ▶ descent in ambient dimension
- ▶ differentiation w.r.t. regular system of parameters

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hybrid approach:

- ▶ descend several dimension steps
(number of steps tunable)
- ▶ use Jacobian criterion for $(I_U, I_{X|U}, g_U)$

The timings

Tabelle : Timings and Memory Usage

	smooth	smoothtst			hybrid			Jacobian	
		time	'parallel'	mem	time	'parallel'	mem	time	mem
$\mathcal{I}_1(6)$	yes	0.24	0.07	0.22	0.18	0.05	0.22	2.5	34
$\mathcal{I}_1(7)$	yes	0.60	0.17	0.24	0.35	0.10	0.22	310	1300
$\mathcal{I}_1(8)$	yes	0.86	0.22	0.32	0.64	0.15	0.23	—	> 20000
$\mathcal{I}_2(3)$	yes	0.22	0.04	0.14	0.08	0.02	0.14	0.05	4.2
$\mathcal{I}_2(4)$	yes	160	9.1	27	40	4.9	190	15	450
$\mathcal{I}_2(5)$	yes	—	—	—	1200	14	510	4000	16000
$\mathcal{I}_3(4)$	no	0.30	0.05	0.22	0.15	0.03	0.22	1.0	8.6
$\mathcal{I}_3(5)$	yes	0.72	0.10	0.22	0.38	0.07	0.22	110	300
$\mathcal{I}_3(6)$	yes	1.3	0.18	0.22	0.83	0.11	0.22	2500	2300
$\mathcal{I}_4(6, 3)$	no	0.02	0.01	0.14	0.02	0.01	0.14	3.1	34
$\mathcal{I}_4(7, 3)$	no	0.04	0.02	0.14	0.04	0.01	0.14	1600	4000
$\mathcal{I}_4(7, 4)$	no	0.10	0.02	0.14	0.10	0.02	0.14	4300	4000

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In general:

- ▶ parallel methods based on charts well-known in algebraic geometry
- ▶ up to now rarely exploited for parallel algorithms

Thank you