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Computing Minimal Polynomials

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The Problem An example Order polynomials

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A Monte Carlo approach Computing order polynomials A Monte Carlo algorithm Back to the example All of this is joint work with Cheryl Praeger

and is based on earlier ideas of

Peter Neumann and Cheryl Praeger.

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The Problem

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An example

Baby Monster group $B = \langle a, b \rangle$ with $a, b \in \mathbb{F}_2^{4370 \times 4370}$ Consider $M := a + b + ab \in \mathbb{F}_2^{4370 \times 4370}$

Computing

the characteristic polynomial χ_M of *M* takes 8.5*s*the minimal polynomial μ_M of *M* takes 9600*s*(times in GAP, other systems behave similarly).

Questions

What is going on here?

What can we do about this?

Is this a typical example?

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Order polynomials

Definition (Order polynomial)

F field, A f.d. F-algebra, $V \in \text{mod-}A$, $v \in V$, $M \in A$. Then the order polynomial $q := \text{ord}_M(v) \in \mathbb{F}[x]$ is the monic polynomial of least degree such that $v \cdot q(M) = 0$.

Definition (Relative order polynomial)

If additionally W < V is *M*-invariant, then we call $\operatorname{ord}_M(v + W)$ the relative order polynomial of $v + W \in V/W$.

Lemma (Generator of annihilator)

The order polynomial $\operatorname{ord}_M(v)$ divides every polynomial $q \in \mathbb{F}[x]$ with $v \cdot q(M) = 0$.

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The standard approach

What is going on here?

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The characteristic polynomial

Let $v_1, \ldots, v_i \in V$, and $V_i := \langle v_1, \ldots, v_i \rangle_M$ the $\mathbb{F}[M]$ -span. Find smallest $d_1 \in \mathbb{N}$ such that $(v_1, v_1 M, v_1 M^2, \ldots, v_1 M^{d_1})$ is linearly dependent. If

$$v_1 M^{d_1} = \sum_{i=0}^{d_1-1} a_i v_1 M^i$$
 then $\operatorname{ord}_M(v_1) = x^{d_1} - \sum_{i=0}^{d_1-1} a_i x^i$.

Choose some $v_2 \in V \setminus \langle v_1 \rangle_M$ and find smallest $d_2 \in \mathbb{N}$, such that $(v_1, v_1 M, \dots, v_1 M^{d_1-1}, v_2, v_2 M, \dots, v_2 M^{d_2})$ is linearly dependent. If

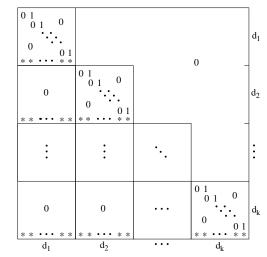
$$v_2 M^{d_2} = \sum_{i=0}^{d_1-1} b_i v_1 M^i + \sum_{i=0}^{d_2-1} c_i v_2 M^i$$
 then
 $\operatorname{ord}_M(v + \langle v_1 \rangle_M) = x^{d_2} - \sum_{i=0} c_i x^i.$

Going on like this we find an \mathbb{F} -basis *Y* of *V*:

$$Y := (v_1, v_1 M, \ldots, v_1^{d_1-1}, \ldots, v_k, v_k M, \ldots, v_k M_k^{d_k-1}).$$

The matrix $Y \cdot M \cdot Y^{-1}$





- Block lower-triangular
- with companion matrices along diagonal
- some sparse garbage below the diagonal

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The minimal polynomial

→ compute the absolute order polynomials $\operatorname{ord}_M(v_i)$ instead the relative ones $\operatorname{ord}_M(v_i + \langle v_1, \dots, v_{i-1} \rangle)_M$.

Lemma (Minimal polynomial)

If $V = \langle v_1, \dots, v_k \rangle_M$ then $\mu_M = \operatorname{lcm}(\operatorname{ord}_M(v_1), \dots, \operatorname{ord}_M(v_k)).$

Problem:

- dim_{\mathbb{F}}(V_i) dim_{\mathbb{F}}(V_{i-1}) might be small
- even if dim_{\mathbb{F}}(V_i) is big.

(set $V_i := \langle v_1, \ldots, v_i \rangle_M$)

Characteristic polynomial: asymptotically $\leq 5n^3$ field ops. Minimal polynomial: asymptotically $\sim n^4$ field ops. (both worst case analysis)

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What can we do about it?

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Two lemmas

Lemma (Order polynomials in cyclic spaces)

Let $W := \langle v \rangle_M < V$ be a cyclic subspace and $p := \operatorname{ord}_M(v)$ be the order polynomial of v. Let $w = v \cdot q(M) \in W$ with deg $(q) < \operatorname{deg}(p)$. Then

$$\operatorname{ord}_M(w) = \frac{p}{\operatorname{gcd}(p,q)}.$$

Lemma (Relative and absolute order polynomials)

Let W < V be *M*-invariant and $v \in V$. If $q := \operatorname{ord}_M(v + W)$ is the relative order polynomial of v, then $v \cdot q(M) \in W$ and

 $\operatorname{ord}_M(v) = q \cdot \operatorname{ord}_M(v \cdot q(M)).$

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Computing order polynomials

We now use the filtration

$$0 = V_0 < V_1 < V_2 < \cdots < V_k = V_1$$

Start with $v \in V_j$ for some $1 \le j \le k$. Then

- compute $q_j := \operatorname{ord}_M(v + V_{j-1})$ in V_j/V_{j-1} (gcd computation with $\operatorname{ord}_M(v_j + V_{j-1})$),
- evaluate $v_j \cdot q_j(M) \in V_{j-1}$,
- proceed inductively,
- take product $\prod_{i=1}^{j} q_i$.

→ use sparseness of YMY^{-1} by "thinking in basis Y" Needs $\leq (j+8) \cdot D^2 + j \cdot D$ field ops. where $D := \dim_{\mathbb{F}}(V_j)$.

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Proposition

Let $\mathbb{F} = \mathbb{F}_q$, randomise $v_1, \ldots, v_u \in V$ independently and uniformly distributed, $\chi_M = \prod_{i=1}^t q_i^{e_i}$. Then:

 $\mathsf{Prob}\left(\mathsf{lcm}(\mathsf{ord}_M(v_1),\ldots,\mathsf{ord}_M(v_u))=\mu_M\right)$

is at least $\prod_{i=1}^{t} (1 - q^{-u \deg(q_i)}).$

Algorithm: Input *M*, $0 < \epsilon < 1/2$

- Compute χ_M , Y, $\operatorname{ord}_M(v_i + V_{i-1})$ for $1 \le i \le k$
- Determine least u, such that probability $> 1 \epsilon$
- Compute $\operatorname{ord}_M(v_1), \ldots, \operatorname{ord}_M(v_u)$
- Return least common multiple

Needs asymptotically $\leq 5n^3 + FACTORISATION(n)$ field ops.

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The new algorithm needs

• 13.3 s to compute μ_M with $\epsilon = 1/100$

30.0 s with deterministic verification afterwards

How typical is this example?

Irreducible factors of χ_M :

deg	1	1	2	4	6	88	197	854	934
χ_{M}	2	2277	4	1	1	1	1	1	1
μ_M	1	5	4	1	1	1	1	1	1

What we see is

- typical behaviour for such matrices,
- most matrices are not of this type,
- however, such matrices might occur in applications.