

The Problem

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The standard
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The characteristic
polynomial
The minimal polynomial

A Monte Carlo
approach

Computing order
polynomials
A Monte Carlo algorithm
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Computing Minimal Polynomials

Max Neunhoffer

Lehrstuhl D für Mathematik
RWTH Aachen

Oberwolfach in July 2006

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

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

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Computing

- the characteristic polynomial χ_M of M takes

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- the characteristic polynomial χ_M of M takes 8.5s

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 8.5s
- the minimal polynomial μ_M of M takes

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 **8.5s**
- the minimal polynomial μ_M of M takes **9600s**

(times in GAP, other systems behave similarly).

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Questions

What is going on here?

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(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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Definition (Order polynomial)

\mathbb{F} field, \mathcal{A} f.d. \mathbb{F} -algebra, $V \in \text{mod-}\mathcal{A}$, $v \in V$, $M \in \mathcal{A}$.

Then the **order polynomial** $q := \text{ord}_M(v) \in \mathbb{F}[x]$ is the monic polynomial of least degree such that $v \cdot q(M) = 0$.

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Definition (Relative order polynomial)

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

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

Lemma (Generator of annihilator)

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

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What is going on here?

The characteristic polynomial

Let $v_1, \dots, v_i \in V$, and $V_i := \langle v_1, \dots, v_i \rangle_M$ the $\mathbb{F}[M]$ -span.

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Let $v_1, \dots, v_i \in V$, and $V_i := \langle v_1, \dots, v_i \rangle_M$ the $\mathbb{F}[M]$ -span.

Find **smallest** $d_1 \in \mathbb{N}$ such that $(v_1, v_1 M, v_1 M^2, \dots, v_1 M^{d_1})$ is linearly dependent.

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

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Choose some $v_2 \in V \setminus \langle v_1 \rangle_M$ and find **smallest** $d_2 \in \mathbb{N}$,
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linearly dependent. If

$$v_2 M^{d_2} = \sum_{i=0}^{d_1-1} b_i v_1 M^i + \sum_{i=0}^{d_2-1} c_i v_2 M^i \quad \text{then}$$

$$\text{ord}_M(v + \langle v_1 \rangle_M) = x^{d_2} - \sum_{i=0}^{d_2-1} c_i x^i.$$

The characteristic polynomial

Let $v_1, \dots, v_i \in V$, and $V_i := \langle v_1, \dots, v_i \rangle_M$ the $\mathbb{F}[M]$ -span.
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Going on like this we find an \mathbb{F} -basis Y of V :

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

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

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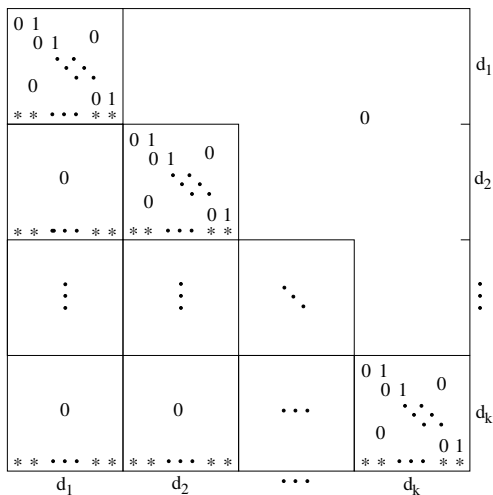
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- Block lower-triangular
- with companion matrices along diagonal
- some sparse garbage below the diagonal

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

The minimal polynomial

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Lemma (Minimal polynomial)

If $V = \langle v_1, \dots, v_k \rangle_M$ then

$$\mu_M = \text{lcm}(\text{ord}_M(v_1), \dots, \text{ord}_M(v_k)).$$

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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 $V_i := \langle v_1, \dots, v_i \rangle_M$)

The minimal polynomial

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

Characteristic polynomial: **asymptotically** $\leq 5n^3$ field ops.

The minimal polynomial

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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?

Lemma (Order polynomials in cyclic spaces)

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

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

Two lemmas

Lemma (Order polynomials in cyclic spaces)

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

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

Lemma (Relative and absolute order polynomials)

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

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

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We now use the filtration

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

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We now use the filtration

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

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

- compute $q_j := \text{ord}_M(v + V_{j-1})$ in V_j/V_{j-1}
(gcd computation with $\text{ord}_M(v_j + V_{j-1})$),

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- evaluate $v_j \cdot q_j(M) \in V_{j-1}$,

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- proceed **inductively**,
- take product $\prod_{i=1}^j q_i$.

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→ use sparseness of YMY^{-1} by “thinking in basis Y ”

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→ use sparseness of YMY^{-1} by “thinking in basis Y ”

Needs $\leq (j+8) \cdot D^2 + j \cdot D$ field ops. where $D := \dim_{\mathbb{F}}(V_j)$.

A Monte Carlo algorithm

Proposition

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

$$\text{Prob}(\text{lcm}(\text{ord}_M(v_1), \dots, \text{ord}_M(v_u)) = \mu_M)$$

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

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

- Compute $\chi_M, Y, \text{ord}_M(v_i + V_{i-1})$ for $1 \leq i \leq k$
- Determine least u , such that probability $> 1 - \epsilon$
- Compute $\text{ord}_M(v_1), \dots, \text{ord}_M(v_u)$
- **Return** least common multiple

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

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

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How typical is this example?

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- **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

Back to the example

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

Consider $M := a + b + ab \in \mathbb{F}_2^{4370 \times 4370}$

The new algorithm needs

- **13.3 s** to compute μ_M with $\epsilon = 1/100$
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- most matrices are **not** of this type,
- however, such matrices might occur in applications.