

Faster algorithms for the characteristic polynomial

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ISSAC 2007, Waterloo,
July 30

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Compute the **characteristic polynomial** of a **dense** matrix over a **field**

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- Improves previous complexity by a $\log n$ factor,
- Optimal reduction to Matrix multiplication.
- Practical efficiency. E.g. over $\mathbb{Z}_{547\,909}$:

n	500	5000	15 000
LinBox	0.91s	4m44s	2h20m
magma-2.13	1.27s	15m32s	7h28m

Outline

- 1 State of the art
- 2 A new algorithm
 - Shifted forms
 - Principle of the new algorithm
 - Complexity
- 3 The new algorithm into practice

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Pre-Strassen age

Leverrier 1840: trace of powers of A , and Newton's formula

- improved/rediscovered by Souriau, Faddeev, Frame and Csanky
- $\mathcal{O}(n^4)$, based on Matrix multiplication
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$$\Rightarrow \mathcal{O}(n^3)$$

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Hessenberg 1942: transformation to quasi-upper triangular and determinant expansion formula.

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Preparata & Sarwate 1978: Update Csanky with fast matrix multiplication

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Keller-Gehrig 1985, alg.2: inspired by Danilevskii, block operations

- $\mathcal{O}(n^\omega)$
- but only valid with generic matrices

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Definition (degree d Krylov matrix of one vector v)

$$K = [v \quad Av \quad \dots \quad A^{d-1}v]$$

Property

$$A \times K = K \times \underbrace{\begin{bmatrix} 0 & & & * \\ 1 & & & * \\ & \ddots & & * \\ & & 1 & * \end{bmatrix}}_{C_{P^{A,v}}^{\min}}$$

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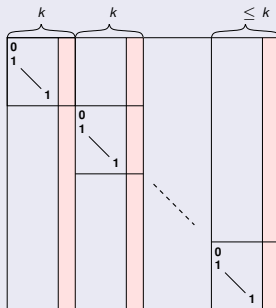
[Keller-Gehrig, alg. 2] : $K^{-1}AK$ in $\mathcal{O}(n^\omega)$ for A generic

Definition (degree k Krylov matrix of several vectors v_i)

$$K = [v_1 \quad \dots \quad A^{k-1} v_1 \mid v_2 \quad \dots \quad A^{k-1} v_2 \mid \dots \mid v_l \quad \dots \quad A^{k-1} v_l]$$

Property

$$A \times K = K \times$$



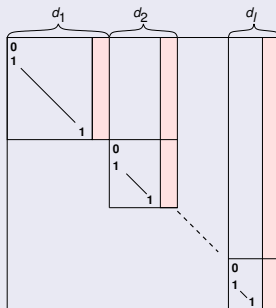
Fact (Shift Hessenberg form)

If (d_1, \dots, d_l) is lexicographically maximal such that

$$K = [v_1 \quad \dots \quad A^{d_1-1} v_1 \mid \dots \mid v_l \quad \dots \quad A^{d_l-1} v_l]$$

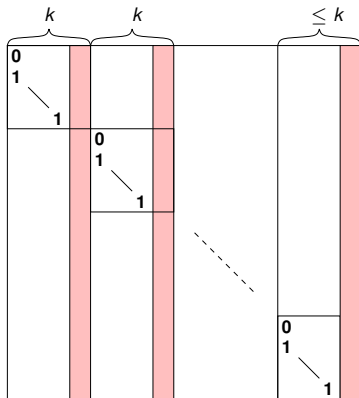
is non-singular, then

$$A \times K = K \times$$



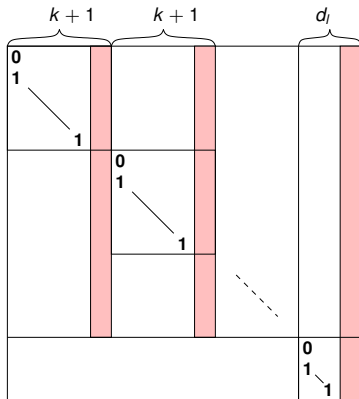
Principle

k -shifted form:



Principle

$k + 1$ -shifted form:



Principle

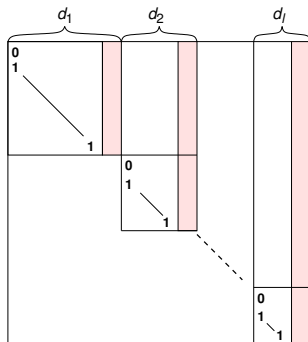
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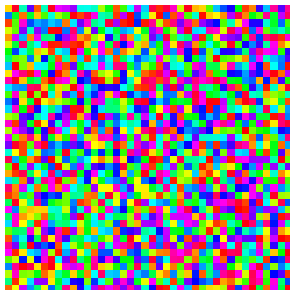
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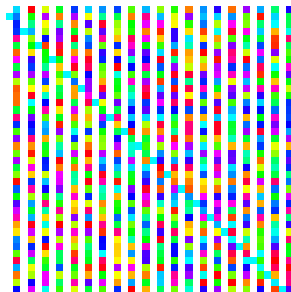
- Compute iteratively from 1-shifted form to d_1 -shifted form
- each completed block appears in the increasing degree order
- until the shifted Hessenberg form is obtained:



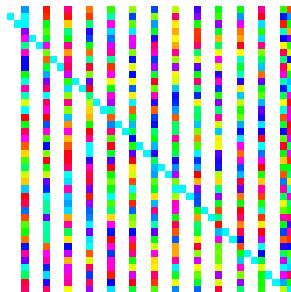
Example



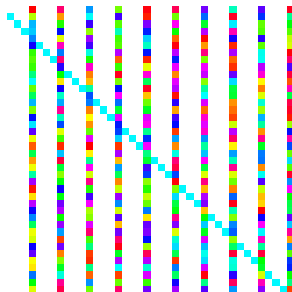
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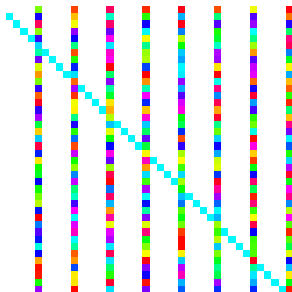
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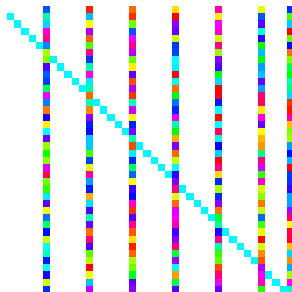
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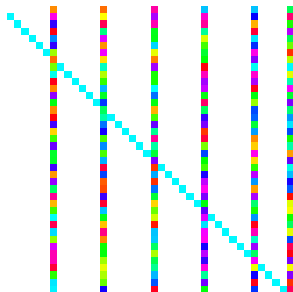
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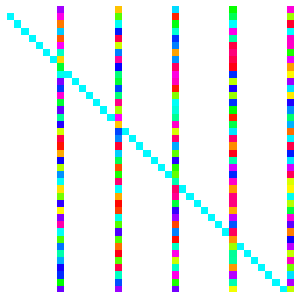
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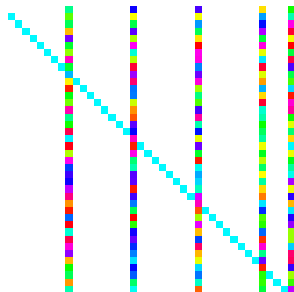
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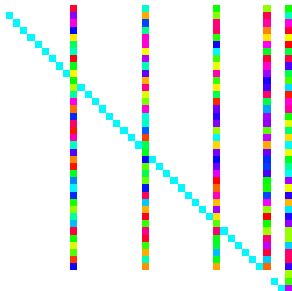
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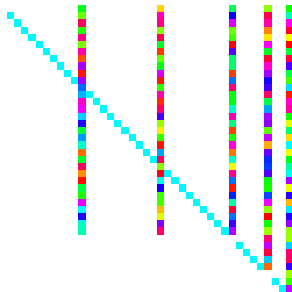
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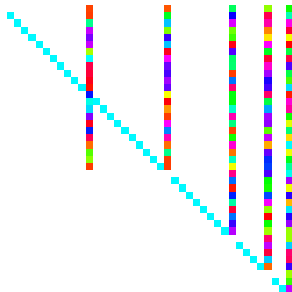
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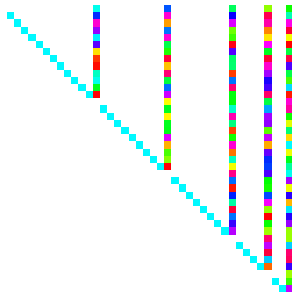
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If $\#F > 2n^2$, the transformation will succeed with high probability. Failure is detected.

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How to use fast matrix arithmetic ?

Reduction to Matrix multiplication

Similarity transformation:

$$K^{-1}AK = Q'^T \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} P'^T Q \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} P Q' \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} P'$$

Reduction to Matrix multiplication

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Rank profile: derived from LQUP

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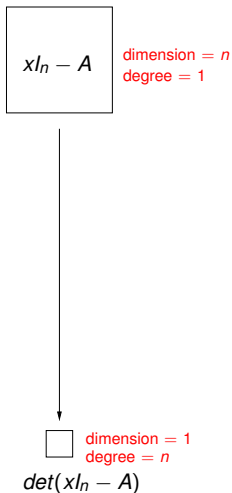
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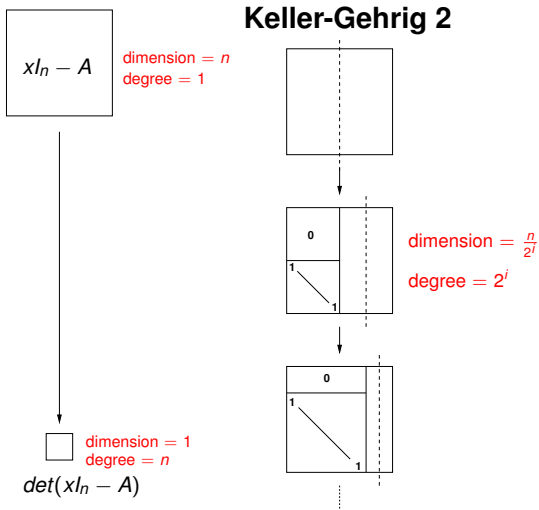
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$$\sum_{k=1}^n k \left(\frac{n}{k}\right)^\omega = n^\omega \sum_{k=1}^n \frac{1}{k^{\omega-1}} = \mathcal{O}(n^\omega)$$

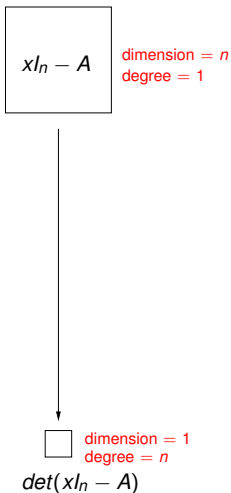
A new type of reduction



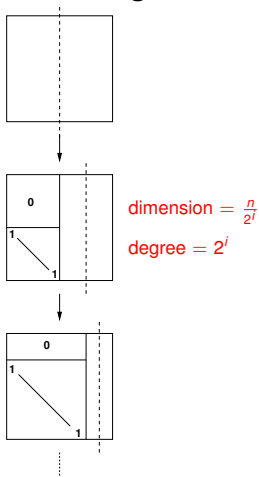
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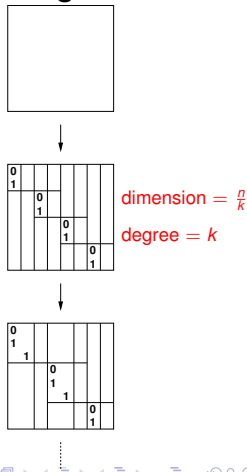
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Keller-Gehrig 2



New algorithm



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Improving the preconditioning

The preconditioning phase:

$$A \leftarrow U^{-1}AU$$

for a random matrix U .

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Instead, use a block Krylov preconditioning:

$$A \leftarrow V^{-1}AV,$$

$$V = [W \quad AW \quad \dots \quad A^{c-1}W]$$

for a random $n \times n/c$ matrix W .

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Property

$V^{-1}AV$ is in c shifted form.

Efficiency balancing parameter

c small: full square matrix multiplications, but more ops

c large: tends to matrix-vector products, but less ops

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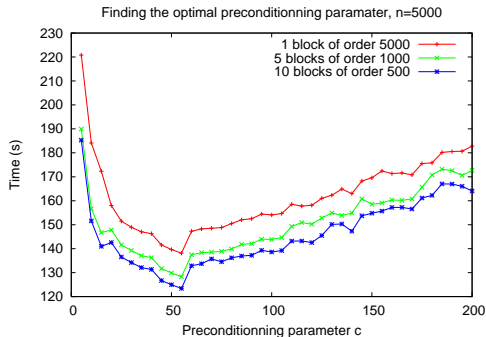
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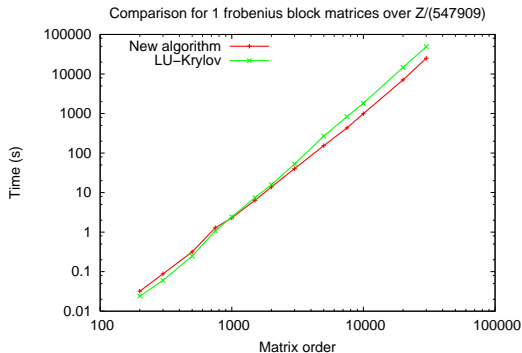


Experiments

n	LU-Krylov	New algorithm
200	0.024	0.032
300	0.06s	0.088s
500	0.248s	0.316s
750	1.084s	1.288s
1000	2.42s	2.296s
5000	267.6s	153.9s
10 000	1827s	991s
20 000	14 652s	7097s
30 000	48 887s	24 928s

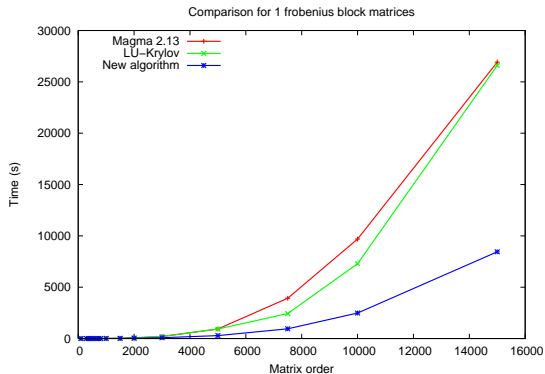
Computation time for 1 Frobenius block matrices, Itanium2-64 1.3Ghz, 192Gb

Experiments



Timing comparison between the new algorithm and LU-Krylov, logarithmic

Comparison to Magma and previous LinBox



Conclusion and perspectives

Results:

- Las Vegas reduction to matrix multiplication,
- The Frobenius normal form is easily derivable in $\mathcal{O}(n^\omega)$...
- ...but no transformation matrix
- Adaptive combination with block Krylov in practice.

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Still to be done:

- Condition on the size of the field is a limitation. Eberly's algorithm ?
- Ideally: derandomization ? (deterministic)
- Unification with matrix polynomial algorithms