

# Linear Algebra 1: Computing canonical forms in exact linear algebra

Clément PERNET,

LIG/INRIA-MOAIS, Grenoble Université, France

ECRYPT II: Summer School on Tools,  
Mykonos, Grece,  
June 1st, 2012

# Introduction : matrix normal forms

Given a transformation  $B = f(A)$ ,

- ▶ Identify invariants
- ▶ Unique representant of an equivalence class
- ▶ Simplify computations (structured form)

Different types:

Equivalence over a field:  $B = UA$ , where  $U$  is invertible

- ▶ Reduced echelon form:

$$E = \begin{bmatrix} 1 & * & 0 & * & * & 0 & * \\ & & 1 & * & * & 0 & * \\ & & & & & 1 & * \end{bmatrix}$$

- ▶ Gauss-Jordan elimination

# Introduction : matrix normal forms

Given a transformation  $B = f(A)$ ,

- ▶ Identify invariants
- ▶ Unique representant of an equivalence class
- ▶ Simplify computations (structured form)

Different types:

Equivalence over a ring:  $B = UA$ , where  $\det(U) = \pm 1$

- ▶ Hermite normal form:

$$0 \leq x_{*,j} < p_j$$

$$H = \begin{bmatrix} p_1 & * & x_{1,2} & * & * & x_{1,3} & * \\ & & p_2 & * & * & x_{2,3} & * \\ & & & & & p_3 & * \end{bmatrix}, \text{ with}$$

# Introduction : matrix normal forms

Given a transformation  $B = f(A)$ ,

- ▶ Identify invariants
- ▶ Unique representant of an equivalence class
- ▶ Simplify computations (structured form)

Different types:

## Similarity over a field: $B = U^{-1}AU$

- ▶ Frobenius normal form (or canonical rational form):

$$F = \begin{bmatrix} C_{P_0} & & & \\ & C_{P_1} & & \\ & & \ddots & \\ & & & C_{P_k} \end{bmatrix}, \text{ with } p_{i+1} | p_i \text{ and } P_0 = \text{MinPoly}(A).$$

- ▶ Krylov method or ZigZag elimination

# Motivation

## Equivalence over a field: Gaussian elimination

- ▶ **Reduced echelon form**, rank profile: PoISys-Gröbner basis.
- ▶ Linear system solving: sieves, index calculus, ...

## Equivalence over a ring: lattice reduction

- ▶ **Hermite normal form**:  $\mathbb{Z}$ -modules and their saturation
- ▶ short vector problem:
  - ▶ hard problem
  - ▶ help improve computational complexities

# Complexities

Matrix multiplication: door to fast linear algebra

- ▶ over a field:  $\mathcal{O}(n^\omega)$ .  $\omega \in ]2.3727, 3]$  (exponent of linear algebra)
- ▶ over  $\mathbb{Z}$ :  $\mathcal{O}(n^\omega M(\log \|A\|)) = \tilde{\mathcal{O}}(n^\omega \log \|A\|)$

Equivalence over a field: Reduced echelon form

- ▶ Gauss-Jordan:  $\mathcal{O}(n^\omega)$

# Complexities

Matrix multiplication: door to fast linear algebra

- ▶ over a field:  $\mathcal{O}(n^\omega)$ .  $\omega \in ]2.3727, 3]$  (exponent of linear algebra)
- ▶ over  $\mathbb{Z}$ :  $\mathcal{O}(n^\omega M(\log \|A\|)) = \tilde{\mathcal{O}}(n^\omega \log \|A\|)$

Equivalence over a field: Reduced echelon form

- ▶ Gauss-Jordan:  $\mathcal{O}(n^\omega)$

Equivalence over  $\mathbb{Z}$ : Hermite normal form

- ▶ [Kannan & Bachem 79]:  $\in P$
- ▶ [Chou & Collins 82]:  $\tilde{\mathcal{O}}(n^6 \log \|A\|)$
- ▶ [Domich & Al. 87], [Illiopoulos 89]:  $\tilde{\mathcal{O}}(n^4 \log \|A\|)$
- ▶ [Micciancio & Warinschi01]:  $\tilde{\mathcal{O}}(n^5 \log \|A\|^2)$ ,
- ▶ [Storjohann & Labahn 96]:  $\mathcal{O}(n^3 \log \|A\|)$  heur.  
 $\tilde{\mathcal{O}}(n^{\omega+1} \log \|A\|)$

# Complexities

Matrix multiplication: door to fast linear algebra

- ▶ over a field:  $\mathcal{O}(n^\omega)$ .  $\omega \in ]2.3727, 3]$  (exponent of linear algebra)
- ▶ over  $\mathbb{Z}$ :  $\mathcal{O}(n^\omega M(\log \|A\|)) = \tilde{\mathcal{O}}(n^\omega \log \|A\|)$

Equivalence over a field: Reduced echelon form

- ▶ Gauss-Jordan:  $\mathcal{O}(n^\omega)$

Equivalence over  $\mathbb{Z}$ : Hermite normal form

- ▶ [Kannan & Bachem 79]:  $\in P$
- ▶ [Chou & Collins 82]:  $\tilde{\mathcal{O}}(n^6 \log \|A\|)$
- ▶ [Domich & Al. 87], [Illiopoulos 89]:  $\tilde{\mathcal{O}}(n^4 \log \|A\|)$
- ▶ [Micciancio & Warinschi01]:  $\tilde{\mathcal{O}}(n^5 \log \|A\|^2)$ ,
- ▶ [Storjohann & Labahn 96]:  $\tilde{\mathcal{O}}(n^3 \log \|A\|)$  heur.  $\tilde{\mathcal{O}}(n^{\omega+1} \log \|A\|)$

Similarity over a field: Frobenius normal form

- ▶ [Storjohann00]:  $\tilde{\mathcal{O}}(n^\omega)$
- ▶ [P. & Storjohann07]: Las Vegas without  $U$   $\mathcal{O}(n^\omega)$



# Complexities

Matrix multiplication: door to fast linear algebra

- ▶ over a field:  $\mathcal{O}(n^\omega)$ .  $\omega \in ]2.3727, 3]$  (exponent of linear algebra)
- ▶ over  $\mathbb{Z}$ :  $\mathcal{O}(n^\omega M(\log \|A\|)) = \tilde{\mathcal{O}}(n^\omega \log \|A\|)$

Equivalence over a field: Reduced echelon form

- ▶ **Gauss-Jordan:**  $\mathcal{O}(n^\omega)$

Equivalence over  $\mathbb{Z}$ : Hermite normal form

- ▶ [Kannan & Bachem 79]:  $\in P$
- ▶ [Chou & Collins 82]:  $\tilde{\mathcal{O}}(n^6 \log \|A\|)$
- ▶ [Domich & Al. 87], [Illiopoulos 89]:  $\tilde{\mathcal{O}}(n^4 \log \|A\|)$
- ▶ **[Micciancio & Warinschi01]:**  $\tilde{\mathcal{O}}(n^5 \log \|A\|^2)$ ,
- ▶ **[Storjohann & Labahn 96]:**  $\tilde{\mathcal{O}}(n^3 \log \|A\|)$  **heur.**  
 $\tilde{\mathcal{O}}(n^{\omega+1} \log \|A\|)$

Similarity over a field: Frobenius normal form

- ▶ [Storjohann00]:  $\tilde{\mathcal{O}}(n^\omega)$
- ▶ **[P. & Storjohann07]: Las Vegas without  $U$**   $\mathcal{O}(n^\omega)$

# Motivation

## Algorithmic building blocks in the design of efficient computation of exact linear algebra normal forms.

### In brief

#### Reductions to a building block

**Matrix Mult:** block rec.  $\sum_{i=1}^{\log n} n \left(\frac{n}{2^i}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Gauss, REF)

**Matrix Mult:** Iterative  $\sum_{k=1}^n n \left(\frac{n}{k}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Frobenius)

**Linear Sys:** over  $\mathbb{Z}$  (Hermite Normal Form)

#### Size/dimension compromises

- ▶ Hermite normal form : rank 1 updates reducing the determinant
- ▶ Frobenius normal form : degree  $k$ , dimension  $n/k$  for  $k = 1 \dots n$

# Motivation

**Algorithmic building blocks in the design of efficient computation of exact linear algebra normal forms.**

## In brief

### Reductions to a building block

**Matrix Mult:** block rec.  $\sum_{i=1}^{\log n} n \left(\frac{n}{2^i}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Gauss, REF)

**Matrix Mult:** Iterative  $\sum_{k=1}^n n \left(\frac{n}{k}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Frobenius)

**Linear Sys:** over  $\mathbb{Z}$  (Hermite Normal Form)

### Size/dimension compromises

- ▶ Hermite normal form : rank 1 updates reducing the determinant
- ▶ Frobenius normal form : degree  $k$ , dimension  $n/k$  for  $k = 1 \dots n$

Study originating from, the design of the libraries

- ▶ FFLAS-FFPACK: dense over word size  $\mathbb{Z}_p$ .
- ▶ M4RI: dense over GF(2) (see Martin Albrecht's Talk)
- ▶ LinBox: dense/sparse/black-box over  $\mathbb{Z}$ ,  $\mathbb{Z}_p$

# Outline

## Reduced Echelon forms and Gaussian elimination

- Gaussian elimination based matrix decompositions

- Relations between decompositions

- Algorithms

## Hermite normal form

- Micciancio & Warinschi algorithm

- Double Determinant

- AddCol

## Frobenius normal form

- Krylov method

- Algorithm

- Reduction to matrix multiplication

# Outline

## Reduced Echelon forms and Gaussian elimination

Gaussian elimination based matrix decompositions

Relations between decompositions

Algorithms

## Hermite normal form

Micciancio & Warinschi algorithm

Double Determinant

AddCol

## Frobenius normal form

Krylov method

Algorithm

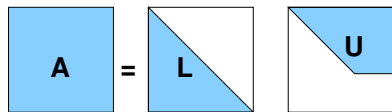
Reduction to matrix multiplication

# Reduced echelon form and Gaussian elimination

## **Gaussian elimination < Reduction to red. Echelon form**

- ▶ Extensively studied for numerical computations
- ▶ Specificities of exact computations:
  - ▶ No partial/full pivoting
  - ▶ Rank profile matters
- ▶ size of coefficients (e.g. compressed in GF(2))  
⇒ asymmetry

# LU decomposition

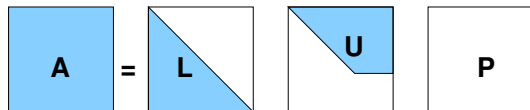


- ▶  $L$  unit lower triangular,
- ▶  $U$  non-sing upper triangular

Exists for

- ▶ matrices having the generic rank profile (every leading principal minor is non zero)

# LUP, PLU decomposition



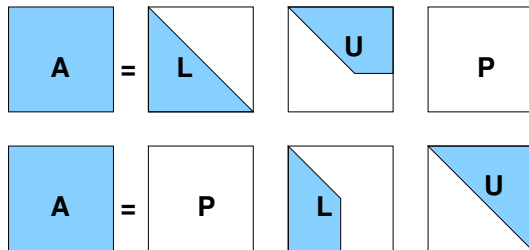
- ▶  $P$  a permutation matrix

Exists for

- ▶ Any non-singular matrix
- ▶ Or any matrix with generic row rank profile



# LUP, PLU decomposition

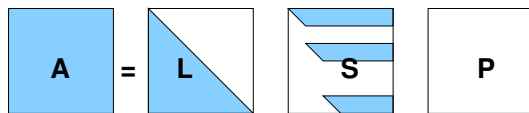


- ▶  $P$  a permutation matrix

Exists for

- ▶ Any non-singular matrix
- ▶ Or any matrix with generic row rank profile

## LSP, LQUP, PLUQ decompositions

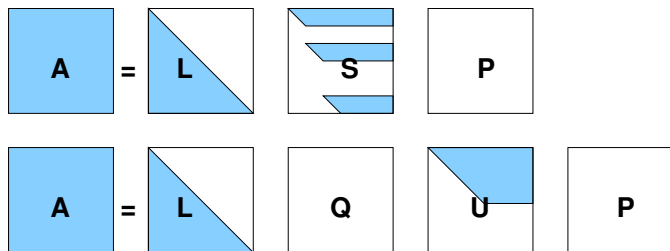


- ▶ **S**: semi-upper triangular,
- ▶ **Q** permutation matrix

Exists for

- ▶ any  $m \times n$  matrix

## LSP, LQUP, PLUQ decompositions

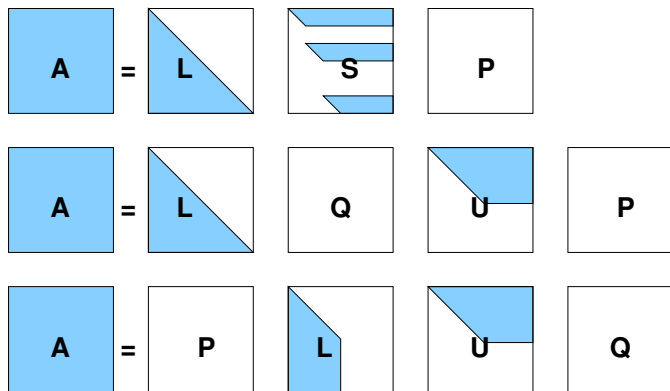


- ▶  $S$ : semi-upper triangular,
- ▶  $Q$  permutation matrix

Exists for

- ▶ any  $m \times n$  matrix

# LSP, LQUP, PLUQ decompositions



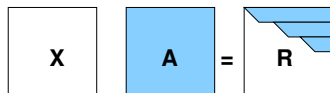
- ▶  $S$ : semi-upper triangular,
- ▶  $Q$  permutation matrix

Exists for

- ▶ any  $m \times n$  matrix

# Echelon form decomposition

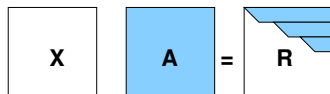
Row Echelon Form  $XA = R$



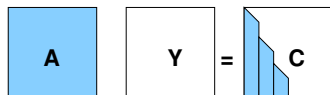
- ▶  $X, Y$ : non-singular transformation matrices
- ▶  $R, C$ : matrices in row/col echelon form

# Echelon form decomposition

Row Echelon Form  $XA = R$



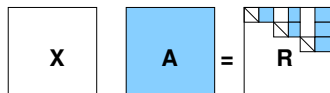
Column Echelon Form  $AY = C$



- ▶  $X, Y$ : non-singular transformation matrices
- ▶  $R, C$ : matrices in row/col echelon form

# Reduced echelon form decomposition

Row Reduced Echelon Form  $XA = R$


$$\boxed{X} \quad \boxed{A} = \boxed{R}$$

- ▶  $X, Y$ : non-singular transformation matrices
- ▶  $R, C$ : matrices in reduced row/col echelon form

# Reduced echelon form decomposition

Row Reduced Echelon Form  $XA = R$

The diagram shows the equation  $XA = R$ . Matrix  $X$  is a white square. Matrix  $A$  is a blue square. Matrix  $R$  is a white square with a blue upper triangular pattern, representing the reduced row echelon form of  $A$ .

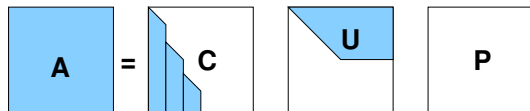
Column Reduced Echelon Form  $AY = C$

The diagram shows the equation  $AY = C$ . Matrix  $A$  is a blue square. Matrix  $Y$  is a white square. Matrix  $C$  is a white square with a blue lower triangular pattern, representing the reduced column echelon form of  $A$ .

- ▶  $X, Y$ : non-singular transformation matrices
- ▶  $R, C$ : matrices in reduced row/col echelon form



# CUP and PLE decompositions

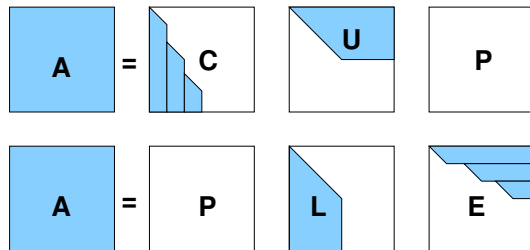


- ▶  $C$ : column echelon form
- ▶  $E$ : row echelon form

Exists for

- ▶ any  $m \times n$  matrix

# CUP and PLE decompositions



- ▶  $C$ : column echelon form
- ▶  $E$ : row echelon form

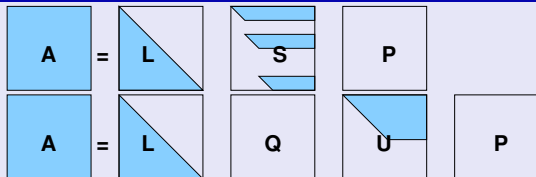
Exists for

- ▶ any  $m \times n$  matrix

# Relations: up to permutations

## From LSP to LQUP

$$S = QU$$



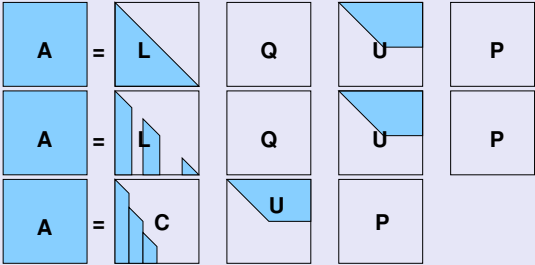
## Fact

*The first  $r = \text{rank}(A)$  values of the permutation  $Q$  are monotonically increasing.*

# Relations: up to permutations

## From LQUP to CUP

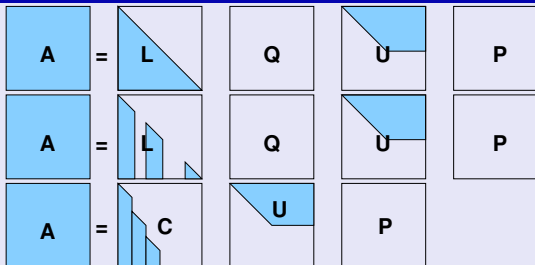
$$C = LQ$$



# Relations: up to permutations

## From LQUP to CUP

$$C = LQ$$



## From LQUP to PLE

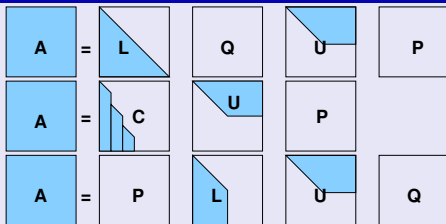
Using transposition:

$$PLE(A^T) = CUP(A)^T$$

# Relations:

## From LQUP to PLUQ

$$P \leftrightarrow Q, L \leftarrow = Q^T L Q$$



# Algorithms: main types

Three ways to group operations:

## 1. simple iterative

- ▶ Apply the standard Gaussian elimination in dimension  $n$
- ▶ Main loop **for**  $i=1$  to  $n$

## 2. block algorithms

### 2.1 block iterative (Tile)

- ▶ Apply Gaussian elimination in dimension  $n/k$  over blocks of size  $k$
- ▶ Main loop: **for**  $i=1$  to  $n/k$

### 2.2 block recursive

- ▶ Apply Gaussian elimination in dimension 2 recursively on blocks of size  $n/2^i$
- ▶ Main loop: **for**  $i=1$  to 2

# Type of algorithms

Data locality: prefer block algorithms

- ▶ cache aware: block iterative
- ▶ cache oblivious: block recursive

Base case efficiency: simple iterative

Asymptotic time complexity: block recursive

Parallelization: block iterative



## Block recursive gaussian elimination

Author	Year	Computation	Requirement
Strassen	69	Inverse	gen. rank prof.
Bunch, Hopcroft	74	LUP	gen. row rank p
Ibarra, Moran, Hui	82	LSP, LQUP	none
Schönage, Keller-Gerig	85	StepForm	none
Storjohann	00	Echelon, RedEch	none
here	11	CUP,PLE,PLUQ	none

## Block recursive gaussian elimination

Author	Year	Computation	Requirement
Strassen	69	Inverse	gen. rank prof.
Bunch, Hopcroft	74	LUP	gen. row rank p
Ibarra, Moran, Hui	82	LSP, LQUP	none
Schönage, Keller-Gerig	85	StepForm	none
Storjohann	00	Echelon, RedEch	none
here	11	CUP,PLE,PLUQ	none

### Comparison according to:

- ▶ No requirement on the input matrix

## Block recursive gaussian elimination

Author	Year	Computation	Requirement
Strassen	69	Inverse	gen. rank prof.
Bunch, Hopcroft	74	LUP	gen. row rank p
Ibarra, Moran, Hui	82	LSP, LQUP	none
Schönage, Keller-Gerig	85	StepForm	none
Storjohann	00	Echelon, RedEch	none
here	11	CUP,PLE,PLUQ	none

### Comparison according to:

- ▶ No requirement on the input matrix
- ▶ Rank sensitive complexity

## Block recursive gaussian elimination

Author	Year	Computation	Requirement
Strassen	69	Inverse	gen. rank prof.
Bunch, Hopcroft	74	LUP	gen. row rank p
Ibarra, Moran, Hui	82	LSP, LQUP	none
Schönage, Keller-Gerig	85	StepForm	none
Storjohann	00	Echelon, RedEch	none
here	11	CUP,PLE,PLUQ	none

### Comparison according to:

- ▶ No requirement on the input matrix
- ▶ Rank sensitive complexity
- ▶ Memory allocations
- ▶ Constant factor in the time complexity

## Memory requirements:

### Definition

In place = *output overrides the input and computation does not need extra memory (considering Matrix multiplication  $C \leftarrow C + AB$  as a black box)*

Remark: a unit lower triangular and an upper triangular matrix can be stored on the same  $m \times n$  storage!

# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

Compute  $F = C^{-1}E$

(Recursive call)

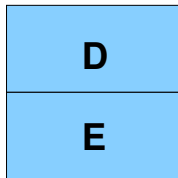
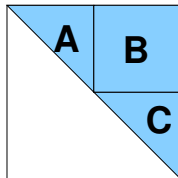
Compute  $G = D - BF$

(MM)

Compute  $H = A^{-1}G$

(Recursive call)

Return  $\begin{bmatrix} H \\ F \end{bmatrix}$



# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

Compute  $F = C^{-1}E$

(Recursive call)

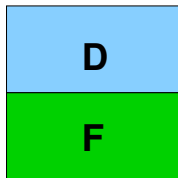
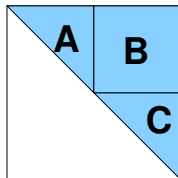
Compute  $G = D - BF$

(MM)

Compute  $H = A^{-1}G$

(Recursive call)

Return  $\begin{bmatrix} H \\ F \end{bmatrix}$





# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

Compute  $F = C^{-1}E$

(Recursive call)

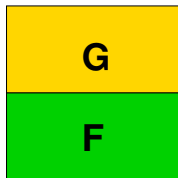
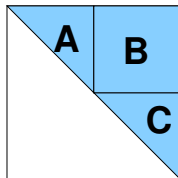
Compute  $G = D - BF$

(MM)

Compute  $H = A^{-1}G$

(Recursive call)

Return  $\begin{bmatrix} H \\ F \end{bmatrix}$



# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

Compute  $F = C^{-1}E$

(Recursive call)

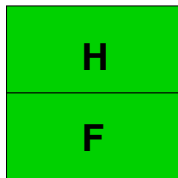
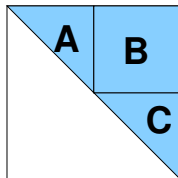
Compute  $G = D - BF$

(MM)

Compute  $H = A^{-1}G$

(Recursive call)

Return  $\begin{bmatrix} H \\ F \end{bmatrix}$



# Preliminaries

## TRSM: TRIangular Solve with Matrix

$$\begin{bmatrix} A & B \\ & C \end{bmatrix}^{-1} \begin{bmatrix} D \\ E \end{bmatrix} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & C^{-1} \end{bmatrix} \begin{bmatrix} D \\ E \end{bmatrix}$$

Compute  $F = C^{-1}E$

(Recursive call)

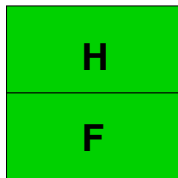
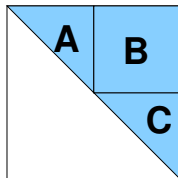
Compute  $G = D - BF$

(MM)

Compute  $H = A^{-1}G$

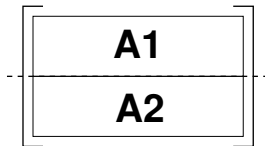
(Recursive call)

Return  $\begin{bmatrix} H \\ F \end{bmatrix}$



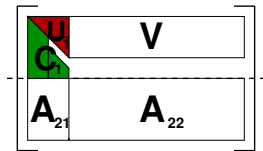
- ▶  $\mathcal{O}(n^\omega)$
- ▶ In place

# The CUP decomposition



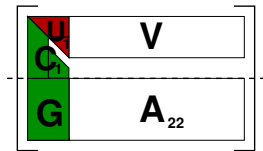
1. Split  $A$  Row-wise

# The CUP decomposition



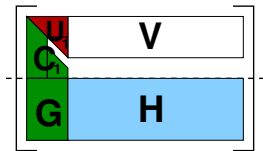
1. Split  $A$  Row-wise
2. Recursive call on  $A_1$

# The CUP decomposition



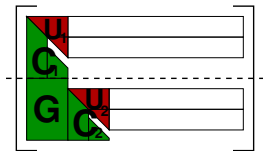
1. Split  $A$  Row-wise
2. Recursive call on  $A_1$
3.  $G \leftarrow A_{21}U_1^{-1}$  (trsm)

# The CUP decomposition



1. Split  $A$  Row-wise
2. Recursive call on  $A_1$
3.  $G \leftarrow A_{21}U_1^{-1}$  (trsm)
4.  $H \leftarrow A_{22} - G \times V$  (MM)

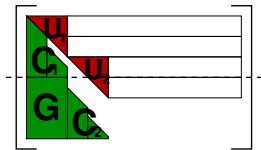
# The CUP decomposition



1. Split  $A$  Row-wise
2. Recursive call on  $A_1$
3.  $G \leftarrow A_{21}U_1^{-1}$  (trsm)
4.  $H \leftarrow A_{22} - G \times V$  (MM)
5. Recursive call on  $H$



# The CUP decomposition



1. Split  $A$  Row-wise
2. Recursive call on  $A_1$
3.  $G \leftarrow A_{21}U_1^{-1}$  (trsm)
4.  $H \leftarrow A_{22} - G \times V$  (MM)
5. Recursive call on  $H$
6. Row permutations

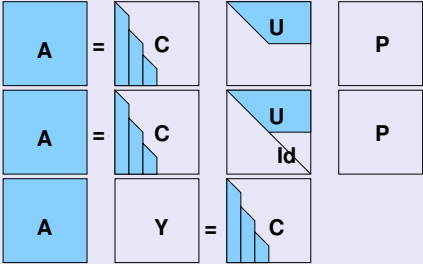
## Memory: LSP vs LQUP vs PLUQ vs CUP

Decomposition	In place
LSP	N
LQUP	N
PLUQ	Y
CUP	Y

# Echelon forms

## From CUP to ColumnEchelon form

$$\begin{aligned}
 Y &= P^T \begin{bmatrix} U_1 & U_2 \\ & I_{n-r} \end{bmatrix}^{-1} \\
 &= P^T \begin{bmatrix} U_1^{-1} & -U_1^{-1}U_2 \\ & I_{n-r} \end{bmatrix}
 \end{aligned}$$



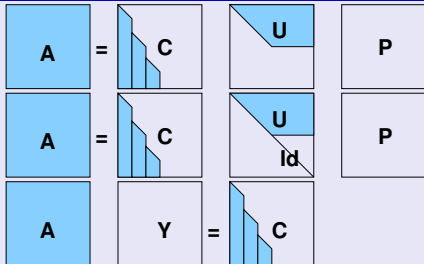
Additional operations:

$-U^{-1}U_2$  `trsm` (triangular system solve) **in-place**

# Echelon forms

## From CUP to ColumnEchelon form

$$\begin{aligned} Y &= P^T \begin{bmatrix} U_1 & U_2 \\ & I_{n-r} \end{bmatrix}^{-1} \\ &= P^T \begin{bmatrix} U_1^{-1} & -U_1^{-1}U_2 \\ & I_{n-r} \end{bmatrix} \end{aligned}$$



Additional operations:

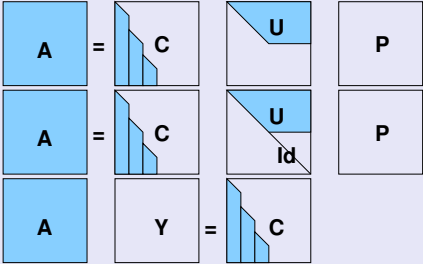
$-U^{-1}U_2$  `trsm` (triangular system solve) **in-place**

$U_1^{-1}$ : `trtri` (triangular inverse)

# Echelon forms

## From CUP to ColumnEchelon form

$$\begin{aligned}
 Y &= P^T \begin{bmatrix} U_1 & U_2 \\ & I_{n-r} \end{bmatrix}^{-1} \\
 &= P^T \begin{bmatrix} U_1^{-1} & -U_1^{-1}U_2 \\ & I_{n-r} \end{bmatrix}
 \end{aligned}$$



Additional operations:

$-U^{-1}U_2$  `trsm` (triangular system solve) **in-place**

$U_1^{-1}$ : `trtri` (triangular inverse) **in-place**

# From CUP to Column Echelon form

## TRTRI: triangular inverse

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix}^{-1} = \begin{bmatrix} U_1^{-1} & -U_1^{-1}U_2U_3^{-1} \\ & U_3^{-1} \end{bmatrix}$$

1: **if**  $n = 1$  **then**

2:  $U \leftarrow U^{-1}$

3: **else**

4:  $U_2 \leftarrow U_3^{-1}U_2$

TRSM

5:  $U_2 \leftarrow -U_2U_3^{-1}$

TRSM

6:  $U_1 \leftarrow U_1^{-1}$

TRTRI

7:  $U_3 \leftarrow U_3^{-1}$

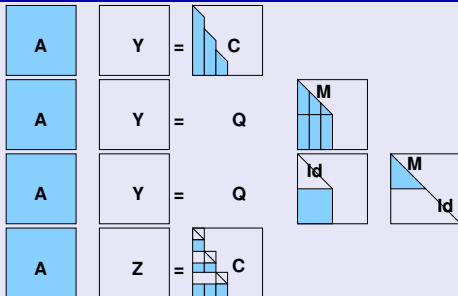
TRTRI

8: **end if**

# Reduced Echelon forms

## From Col. Echelon form to Reduced Col. Echelon form

$$Z = Y \begin{bmatrix} M & \\ & I_{n-r} \end{bmatrix}^{-1}$$



Similarly, from PLE to Row Echelon form

Again reduces to:

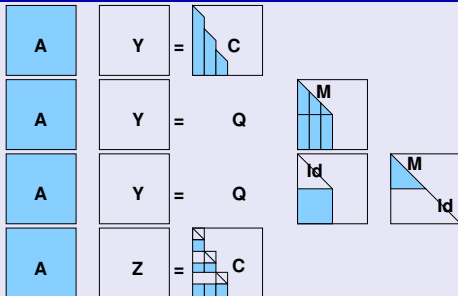
$U^{-1}X$ : TRSM, **in-place**

$U^{-1}$ : TRTRI, **in-place**

# Reduced Echelon forms

## From Col. Echelon form to Reduced Col. Echelon form

$$Z = Y \begin{bmatrix} M & \\ & I_{n-r} \end{bmatrix}^{-1}$$



Similarly, from PLE to Row Echelon form

Again reduces to:

$U^{-1}X$ : TRSM, **in-place**

$U^{-1}$ : TRTRI, **in-place**

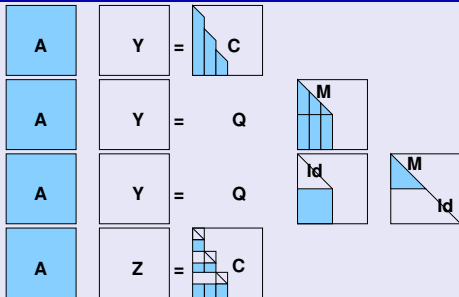
$UL$ : TRTRM,



# Reduced Echelon forms

## From Col. Echelon form to Reduced Col. Echelon form

$$Z = Y \begin{bmatrix} M & \\ & I_{n-r} \end{bmatrix}^{-1}$$



Similarly, from PLE to Row Echelon form

Again reduces to:

$U^{-1}X$ : TRSM, **in-place**

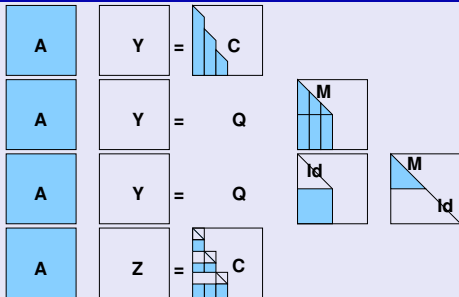
$U^{-1}$ : TRTRI, **in-place**

$UL$ : TRTRM,

# Reduced Echelon forms

## From Col. Echelon form to Reduced Col. Echelon form

$$Z = Y \begin{bmatrix} M & \\ & I_{n-r} \end{bmatrix}^{-1}$$



Similarly, from PLE to Row Echelon form

Again reduces to:

$U^{-1}X$ : TRSM, **in-place**

$U^{-1}$ : TRTRI, **in-place**

$UL$ : TRTRM, **in-place**

# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

- |                                  |       |
|----------------------------------|-------|
| 1: $X_1 \leftarrow U_1L_1$       | TRTRM |
| 2: $X_1 \leftarrow X_1 + U_2L_2$ | MM    |
| 3: $X_2 \leftarrow U_2L_3$       | TRMM  |
| 4: $X_3 \leftarrow U_3L_2$       | TRMM  |
| 5: $X_4 \leftarrow U_3L_3$       | TRTRM |

# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

1:  $X_1 \leftarrow U_1L_1$

TRTRM

2:  $X_1 \leftarrow X_1 + U_2L_2$

MM

3:  $X_2 \leftarrow U_2L_3$

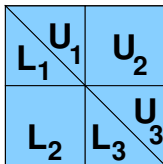
TRMM

4:  $X_3 \leftarrow U_3L_2$

TRMM

5:  $X_4 \leftarrow U_3L_3$

TRTRM

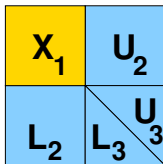


# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

- |                                  |       |
|----------------------------------|-------|
| 1: $X_1 \leftarrow U_1L_1$       | TRTRM |
| 2: $X_1 \leftarrow X_1 + U_2L_2$ | MM    |
| 3: $X_2 \leftarrow U_2L_3$       | TRMM  |
| 4: $X_3 \leftarrow U_3L_2$       | TRMM  |
| 5: $X_4 \leftarrow U_3L_3$       | TRTRM |

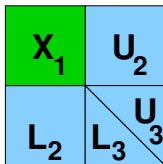


# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

- |                                  |       |
|----------------------------------|-------|
| 1: $X_1 \leftarrow U_1L_1$       | TRTRM |
| 2: $X_1 \leftarrow X_1 + U_2L_2$ | MM    |
| 3: $X_2 \leftarrow U_2L_3$       | TRMM  |
| 4: $X_3 \leftarrow U_3L_2$       | TRMM  |
| 5: $X_4 \leftarrow U_3L_3$       | TRTRM |



# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

1:  $X_1 \leftarrow U_1L_1$

TRTRM

2:  $X_1 \leftarrow X_1 + U_2L_2$

MM

3:  $X_2 \leftarrow U_2L_3$

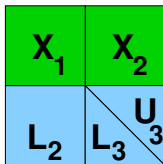
TRMM

4:  $X_3 \leftarrow U_3L_2$

TRMM

5:  $X_4 \leftarrow U_3L_3$

TRTRM

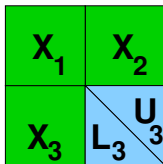


# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

- |                                  |       |
|----------------------------------|-------|
| 1: $X_1 \leftarrow U_1L_1$       | TRTRM |
| 2: $X_1 \leftarrow X_1 + U_2L_2$ | MM    |
| 3: $X_2 \leftarrow U_2L_3$       | TRMM  |
| 4: $X_3 \leftarrow U_3L_2$       | TRMM  |
| 5: $X_4 \leftarrow U_3L_3$       | TRTRM |



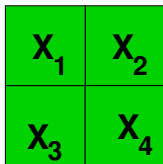


# From Echelon to Reduced Echelon

## TRTRM: triangular triangular multiplication

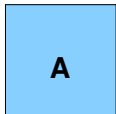
$$\begin{bmatrix} U_1 & U_2 \\ & U_3 \end{bmatrix} \begin{bmatrix} L_1 & \\ L_2 & L_3 \end{bmatrix} = \begin{bmatrix} U_1L_1 + U_2L_2 & U_2L_3 \\ & U_3L_3 \end{bmatrix}$$

- |                                  |       |
|----------------------------------|-------|
| 1: $X_1 \leftarrow U_1L_1$       | TRTRM |
| 2: $X_1 \leftarrow X_1 + U_2L_2$ | MM    |
| 3: $X_2 \leftarrow U_2L_3$       | TRMM  |
| 4: $X_3 \leftarrow U_3L_2$       | TRMM  |
| 5: $X_4 \leftarrow U_3L_3$       | TRTRM |

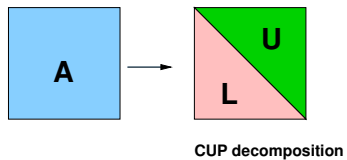


- ▶  $\mathcal{O}(n^\omega)$
- ▶ In place

## Example: in place matrix inversion

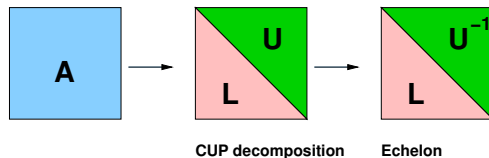


## Example: in place matrix inversion



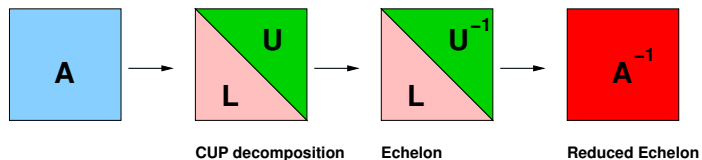
$$A = LU$$

## Example: in place matrix inversion



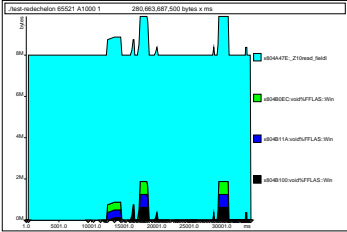
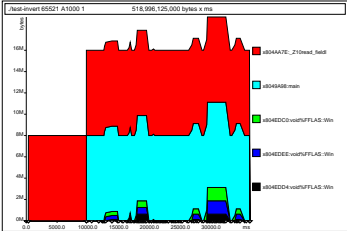
$$AU^{-1} = L$$

## Example: in place matrix inversion



$$A(U^{-1}L^{-1}) = I$$

# Experiments



## Direct computation of the Reduced Echelon form

- ▶ Strassen 69: inverse of generic matrices
- ▶ Storjohann 00: Gauss-Jordan generalization for any rank profile

### Matrix Inversion [Strassen 69]

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & (D - CA^{-1}B)^{-1} \end{bmatrix} \begin{bmatrix} I & \\ CA^{-1} & I \end{bmatrix}$$

# Direct computation of the Reduced Echelon form

- ▶ Strassen 69: inverse of generic matrices
- ▶ Storjohann 00: Gauss-Jordan generalization for any rank profile

## Matrix Inversion [Strassen 69]

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1} = \begin{bmatrix} A^{-1} & \\ & I \end{bmatrix} \begin{bmatrix} I & -B \\ & I \end{bmatrix} \begin{bmatrix} I & \\ & (D - CA^{-1}B)^{-1} \end{bmatrix} \begin{bmatrix} I & \\ CA^{-1} & I \end{bmatrix}$$

- 1: Compute  $E = A^{-1}$  (Recursive call)
- 2: Compute  $F = D - CEB$  (MM)
- 3: Compute  $G = F^{-1}$  (Recursive call)
- 4: Compute  $H = -EB$  (MM)
- 5: Compute  $J = HG$  (MM)
- 6: Compute  $K = CE$  (MM)
- 7: Compute  $L = E + JK$  (MM)
- 8: Compute  $M = GK$  (MM)
- 9: Return  $\begin{bmatrix} E & J \\ M & G \end{bmatrix}$



# Strassen-Storjohann's Gauss-Jordan elimination

## Problem

*Needs to perform operations of the form  $A \leftarrow AB$*

*$\Rightarrow$  not doable in place by a usual matrix multiplication algorithm*

# Strassen-Storjohann's Gauss-Jordan elimination

## Problem

*Needs to perform operations of the form  $A \leftarrow AB$*

*$\Rightarrow$  not doable in place by a usual matrix multiplication algorithm*

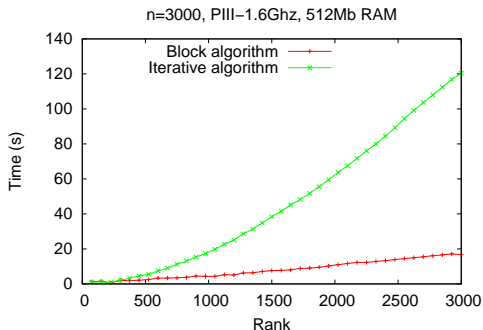
Workaround [Storjohann]:

1. Decompose  $B = LU$  LU
2.  $A \leftarrow AL$  trmm
3.  $A \leftarrow AU$  trmm

# Rank sensitive time complexity

## Fact

Algorithms LSP, CUP, LQUP, PLUQ, ... have a rank sensitive computation time:  $\mathcal{O}(mnr^{\omega-2})$



# Time complexity: comparing constants

$$\mathcal{O}(n^\omega) = C_\omega n^3$$

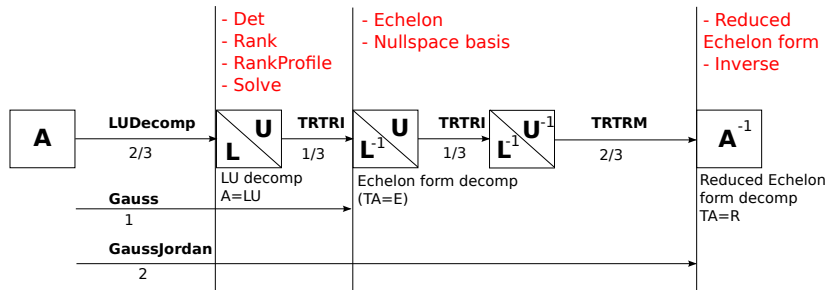
Algorithm	Constant $C_\omega$	$C_3$	$C_{\log_2 7}$	in-place
MM	$C_\omega$	2	6	×
TRSM	$\frac{C_\omega}{2^{\omega-1}-2}$	1	4	✓
TRTRI	$\frac{C_\omega}{(2^{\omega-1}-2)(2^{\omega-1}-1)}$	$\frac{1}{3} \approx 0.33$	$\frac{8}{5} = 1.6$	✓
TRTRM, CUP PLUQ LQUP,	$\frac{C_\omega}{2^{\omega-1}-2} - \frac{C_\omega}{2^{\omega-2}}$	$\frac{2}{3} \approx 0.66$	$\frac{14}{5} = 2.8$	✓
Echelon	$\frac{C_\omega}{2^{\omega-2}-1} - \frac{3C_\omega}{2^{\omega-2}}$	1	$\frac{22}{5} \approx 4.4$	✓
RedEchelon	$\frac{C_\omega(2^{\omega-1}+2)}{(2^{\omega-1}-2)(2^{\omega-1}-1)}$	2	$\frac{44}{5} = 8.8$	✓
StepForm	$\frac{5C_\omega}{2^{\omega-1}-1} + \frac{C_\omega}{(2^{\omega-1}-1)(2^{\omega-2}-1)}$	4	$\frac{76}{5} = 15.2$	×
GJ*	$\frac{C_\omega}{2^{\omega-2}-1}$	2	8	×

\*: GJ: GaussJordan alg of [Storjohann00] computing the reduced echelon form

# Applications to standard linalg problems

Problem	Using	$C_\omega$	$C_3$	$C_{\log_2 7}$	In place
Rank					
RankProfile	GJ	$\frac{C_\omega}{2^{\omega-2}-1}$	2	8	×
IsSingular	CUP	$\frac{C_\omega}{2^{\omega-1}-2} - \frac{C_\omega}{2^{\omega-2}}$	0.66	2.8	✓
Det					
Solve					
Inverse	GJ	$\frac{C_\omega}{2^{\omega-2}-1}$	2	8	×
	CUP	$\frac{C_\omega(2^{\omega-1}+2)}{(2^{\omega-1}-2)(2^{\omega-1}-1)}$	2	8.8	✓

# Summary



# Outline

## Reduced Echelon forms and Gaussian elimination

Gaussian elimination based matrix decompositions

Relations between decompositions

Algorithms

## Hermite normal form

Micciancio & Warinschi algorithm

Double Determinant

AddCol

## Frobenius normal form

Krylov method

Algorithm

Reduction to matrix multiplication

# Computing Hermite Normal form

Equivalence over a ring:  $H = UA$ , where  $\det(U) = \pm 1$

Hermite normal form:  $H = \begin{bmatrix} p_1 & * & x_{1,2} & * & * & x_{1,3} & * \\ & & p_2 & * & * & x_{2,3} & * \\ & & & & & p_3 & * \end{bmatrix}$ , with  $0 \leq x_{*,j} < p_j$

*Reduced Echelon form over a Ring*

## Improving Micciancio-Warinschi algorithm

- ▶  $\mathcal{O}(n^5 \log \|A\|)$  (heuristically:  $\mathcal{O}(n^3 \log \|A\|)$ )
  - ▶ space:  $\mathcal{O}(n^2 \log \|A\|)$
- ⇒ Good on random matrices, common in num. theory, crypto.



# Computing Hermite Normal form

Equivalence over a ring:  $H = UA$ , where  $\det(U) = \pm 1$

Hermite normal form:  $H = \begin{bmatrix} p_1 & * & x_{1,2} & * & * & x_{1,3} & * \\ & & p_2 & * & * & x_{2,3} & * \\ & & & & p_3 & * & * \end{bmatrix}$ , with  $0 \leq x_{*,j} < p_j$

*Reduced Echelon form over a Ring*

## Improving Micciancio-Warinschi algorithm

- ▶  $\mathcal{O} \left( n^5 \log \|A\| \right)$  (heuristically:  $\mathcal{O} \left( n^3 \log \|A\| \right)$ )
- ▶ space:  $\mathcal{O} \left( n^2 \log \|A\| \right)$

⇒ Good on random matrices, common in num. theory, crypto.  
Implementation, reduction to building blocks:

- ▶ LinSys over  $\mathbb{Z}$ ,
- ▶ CUP and MatMul over  $\mathbb{Z}_p$

# Naive algorithm

```
1 begin
2   foreach  $i$  do
3      $(g, t_i, \dots, t_n) = \text{xgcd}(A_{i,i}, A_{i+1,i}, \dots, A_{n,i});$ 
4      $L_i \leftarrow \sum_{j=i+1}^n t_j L_j;$ 
5     for  $j = i + 1 \dots n$  do
6        $L_j \leftarrow L_j - \frac{A_{j,i}}{g} L_i;$           /* eliminate */
7     for  $j = 1 \dots i - 1$  do
8        $L_j \leftarrow L_j - \lfloor \frac{A_{j,i}}{g} \rfloor L_i;$       /* reduce */
9 end
```

$$\begin{bmatrix} p_1 & * & x_{1,2} & * & * & x_{1,3} & * \\ & & p_2 & * & * & x_{2,3} & * \\ & & & & & p_3 & * \end{bmatrix}$$

# Computing modulo the determinant [Domich & Al. 87]

## Property

For  $A$  non-singular:  $\max_i \sum_j H_{ij} \leq \det H$

## Example

$$A = \begin{bmatrix} -5 & 8 & -3 & -9 & 5 & 5 \\ -2 & 8 & -2 & -2 & 8 & 5 \\ 7 & -5 & -8 & 4 & 3 & -4 \\ 1 & -1 & 6 & 0 & 8 & -3 \end{bmatrix}, H = \begin{bmatrix} 1 & 0 & 3 & 237 & -299 & 90 \\ 0 & 1 & 1 & 103 & -130 & 40 \\ 0 & 0 & 4 & 352 & -450 & 135 \\ 0 & 0 & 0 & 486 & -627 & 188 \end{bmatrix}$$

$$\det A = 1944$$

# Computing modulo the determinant [Domich & Al. 87]

## Property

For  $A$  non-singular:  $\max_i \sum_j H_{ij} \leq \det H$

## Example

$$A = \begin{bmatrix} -5 & 8 & -3 & -9 & 5 & 5 \\ -2 & 8 & -2 & -2 & 8 & 5 \\ 7 & -5 & -8 & 4 & 3 & -4 \\ 1 & -1 & 6 & 0 & 8 & -3 \end{bmatrix}, H = \begin{bmatrix} 1 & 0 & 3 & 237 & -299 & 90 \\ 0 & 1 & 1 & 103 & -130 & 40 \\ 0 & 0 & 4 & 352 & -450 & 135 \\ 0 & 0 & 0 & 486 & -627 & 188 \end{bmatrix}$$

$$\det A = 1944$$

Moreover, every computation can be done modulo  $d = \det A$ :

$$U' \begin{bmatrix} A & \\ dI_n & I_n \end{bmatrix} = \begin{bmatrix} H & \\ & I_n \end{bmatrix}$$

$$\Rightarrow \mathcal{O}(n^3) \times M(n(\log n + \log \|A\|)) = \tilde{\mathcal{O}}(n^4 \log \|A\|)$$

## Computing modulo the determinant

- ▶ Pessimistic estimate on the arithmetic size
  - ▶  $d$  large but most coefficients of  $H$  are small
  - ▶ *On the average* : only the last few columns are *large*
- ⇒ Compute  $H'$  close to  $H$  but with small determinant

# Computing modulo the determinant

- ▶ Pessimistic estimate on the arithmetic size
  - ▶  $d$  large but most coefficients of  $H$  are small
  - ▶ *On the average* : only the last few columns are *large*
- ⇒ Compute  $H'$  close to  $H$  but with small determinant

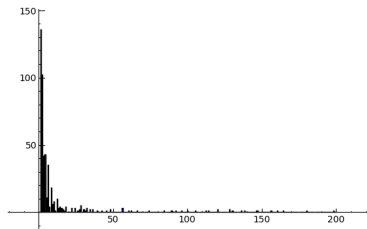
[Micciancio & Warinschi 01]

$$A = \begin{bmatrix} B & b \\ c^T & a_{n-1,n} \\ d^T & a_{n,n} \end{bmatrix}$$

$$d_1 = \det \left( \begin{bmatrix} B \\ c^T \end{bmatrix} \right), d_2 = \det \left( \begin{bmatrix} B \\ d^T \end{bmatrix} \right)$$

$$g = \gcd(d_1, d_2) = sd_1 + td_2 \quad \text{Then}$$

$$\det \left( \begin{bmatrix} B \\ sc^T + td^T \end{bmatrix} \right) = g$$



# Micciancio & Warinschi algorithm

```
1 begin
2   Compute  $d_1 = \det \left( \begin{bmatrix} B \\ c^T \end{bmatrix} \right), d_2 = \det \left( \begin{bmatrix} B \\ d^T \end{bmatrix} \right);$            /* Double Det */
3    $(g, s, t) = \text{xgcd}(d_1, d_2);$ 
4   Compute  $H_1$  the HNF of  $\begin{bmatrix} B \\ sc^T + td^T \end{bmatrix} \pmod{g};$            /* Modular HNF */
5   Recover  $H_2$  the HNF of  $\begin{bmatrix} B & b \\ sc^T + td^T & sa_{n-1,n} + ta_{n,n} \end{bmatrix};$            /* AddCol */
6   Recover  $H_3$  the HNF of  $\begin{bmatrix} B & b \\ c^T & a_{n-1,n} \\ d^T & a_{n,n} \end{bmatrix};$            /* AddRow */
7 end
```

# Micciancio & Warinschi algorithm

```
1 begin
2   Compute  $d_1 = \det \left( \begin{bmatrix} B \\ c^T \end{bmatrix} \right), d_2 = \det \left( \begin{bmatrix} B \\ d^T \end{bmatrix} \right);$  /* Double Det */
3    $(g, s, t) = \text{xgcd}(d_1, d_2);$ 
4   Compute  $H_1$  the HNF of  $\begin{bmatrix} B \\ sc^T + td^T \end{bmatrix} \pmod{g};$  /* Modular HNF */
5   Recover  $H_2$  the HNF of  $\begin{bmatrix} B & b \\ sc^T + td^T & sa_{n-1,n} + ta_{n,n} \end{bmatrix};$  /* AddCol */
6   Recover  $H_3$  the HNF of  $\begin{bmatrix} B & b \\ c^T & a_{n-1,n} \\ d^T & a_{n,n} \end{bmatrix};$  /* AddRow */
7 end
```



# Double Determinant

## First approach: LU mod $p_1, \dots, p_k$ + CRT

- ▶ Only one elimination for the  $n - 2$  first rows
- ▶ 2 updates for the last rows (triangular back substitution)
- ▶  $k$  large such that  $\prod_{i=1}^k p_i > n^n \log \|A\|^{n/2}$

# Double Determinant

## First approach: LU mod $p_1, \dots, p_k$ + CRT

- ▶ Only one elimination for the  $n - 2$  first rows
- ▶ 2 updates for the last rows (triangular back substitution)
- ▶  $k$  *large* such that  $\prod_{i=1}^k p_i > n^n \log \|A\|^{n/2}$

## Second approach: [Abbott Bronstein Mulders 99]

- ▶ Solve  $Ax = b$ .
- ▶  $\delta = \text{lcm}(q_1, \dots, q_n)$  s.t.  $x_i = p_i/q_i$

Then  $\delta$  is a *large* divisor of  $D = \det A$ .

- ▶ Compute  $D/\delta$  by LU mod  $p_1, \dots, p_k$  + CRT
- ▶  $k$  *small*, such that  $\prod_{i=1}^k p_i > n^n \log \|A\|^{n/2} / \delta$

## Double Determinant : improved

### Property

*Let  $x = [x_1, \dots, x_n]$  be the solution of  $[ A \mid c ] x = d$ . Then  $y = [-\frac{x_1}{x_n}, \dots, -\frac{x_{n-1}}{x_n}, \frac{1}{x_n}]$  is the solution of  $[ A \mid d ] y = c$ .*

- ▶ 1 system solve
- ▶ 1 LU for each  $p_i$

$\Rightarrow d_1, d_2$  computed at about the cost of 1 déterminant

# AddCol

## Problem

Find a vector  $e$  such that

$$\left[ H_1 \mid e \right] = U \begin{bmatrix} B & b \\ sc^T + td^T & sa_{n-1,n} + ta_{n,n} \end{bmatrix}$$

$$\begin{aligned} e &= U \begin{bmatrix} b \\ sa_{n-1,n} + ta_{n,n} \end{bmatrix} \\ &= H_1 \begin{bmatrix} B \\ sc^T + td^T \end{bmatrix}^{-1} \begin{bmatrix} b \\ sa_{n-1,n} + ta_{n,n} \end{bmatrix} \end{aligned}$$

⇒ Solve a system.

- ▶  $n - 1$  first rows are *small*
- ▶ last row is *large*

# AddCol

Idea:

replace the last row by a random *small* one  $w^T$ .

$$\begin{bmatrix} B \\ w^T \end{bmatrix} y = \begin{bmatrix} b \\ a_{n-1,n-1} \end{bmatrix}$$

Let  $k$  be a basis of the kernel of  $B$ . Then

$$x = y + \alpha k.$$

where

$$\alpha = \frac{a_{n-1,n-1} - (sc^T + td^T) \cdot y}{(sc^T + td^T) \cdot k}$$

⇒ limits the *expensive* arithmetic to a few dot products

# Outline

## Reduced Echelon forms and Gaussian elimination

Gaussian elimination based matrix decompositions

Relations between decompositions

Algorithms

## Hermite normal form

Micciancio & Warinschi algorithm

Double Determinant

AddCol

## Frobenius normal form

Krylov method

Algorithm

Reduction to matrix multiplication

# Krylov Method

Definition (degree  $d$  Krylov matrix of one vector  $v$ )

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

Property

$$A \times K = K \times \begin{bmatrix} 0 & & & * \\ 1 & & & * \\ & \ddots & & * \\ & & 1 & * \end{bmatrix}$$

# Krylov Method

Definition (degree  $d$  Krylov matrix of one vector  $v$ )

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

Property

$$A \times K = K \times \begin{bmatrix} 0 & & & * \\ 1 & & & * \\ & \ddots & & * \\ & & 1 & * \end{bmatrix}$$

$\Rightarrow$  if  $d = n$ ,

$$K^{-1}AK = C_{P_{car}^A}$$



# Krylov Method

Definition (degree  $d$  Krylov matrix of one vector  $v$ )

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

Property

$$A \times K = K \times \begin{bmatrix} 0 & & & * \\ 1 & & & * \\ & \ddots & & * \\ & & 1 & * \end{bmatrix}$$

$\Rightarrow$  if  $d = n$ ,

$$K^{-1}AK = C_{P^A_{car}}$$

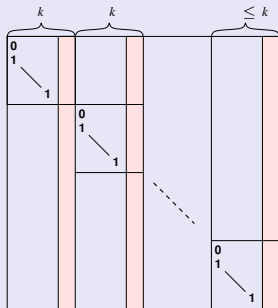
$\Rightarrow$  [Keller-Gehrig, alg. 2] computes  $K$  in  $\mathcal{O}(n^\omega \log n)$

## 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

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



# Hessenberg poly-cyclic form

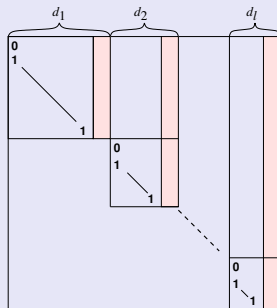
## Fact

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

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

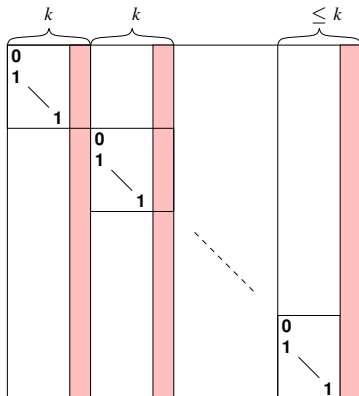
is non-singular, then

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



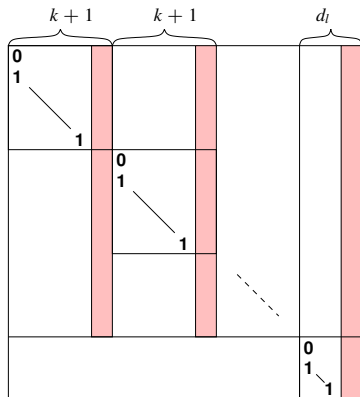
# Principle

$k$ -shifted form:



# Principle

$k + 1$ -shifted form:



## Principle

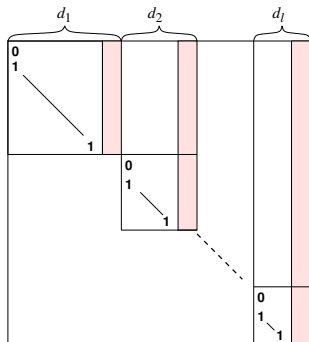
- ▶ Compute iteratively from 1-shifted form to  $d_1$ -shifted form

## Principle

- ▶ Compute iteratively from 1-shifted form to  $d_1$ -shifted form
- ▶ each diagonal block appears in the increasing degree

# Principle

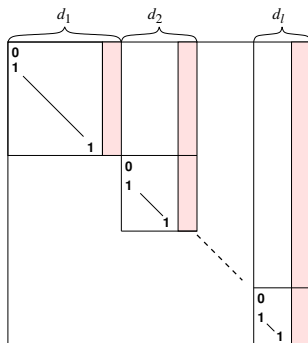
- ▶ Compute iteratively from 1-shifted form to  $d_1$ -shifted form
- ▶ each diagonal block appears in the increasing degree
- ▶ until the shifted Hessenberg form is obtained:





# Principle

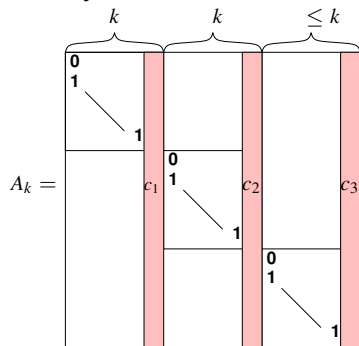
- ▶ Compute iteratively from 1-shifted form to  $d_1$ -shifted form
- ▶ each diagonal block appears in the increasing degree
- ▶ until the shifted Hessenberg form is obtained:



How to transform from  $k$  to  $k + 1$ -shifted form ?

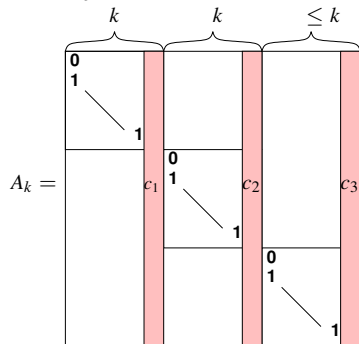
# Krylov normal extension

for any  $k$ -shifted form

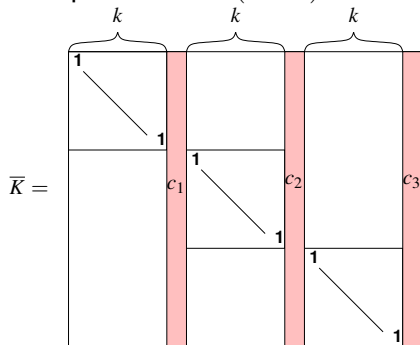


# Krylov normal extension

for any  $k$ -shifted form

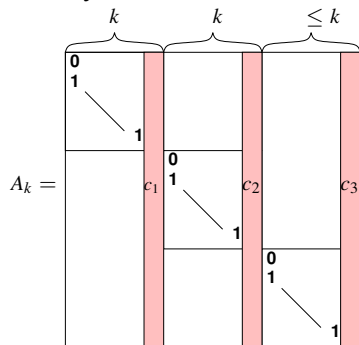


compute the  $n \times (n + k)$  matrix

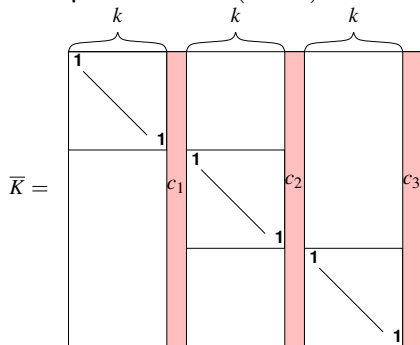


# Krylov normal extension

for any  $k$ -shifted form



compute the  $n \times (n + k)$  matrix



and form  $K$  by picking its first linearly independent columns.

# The algorithm

- ▶ Form  $\bar{K}$ : just copy the columns of  $A_k$

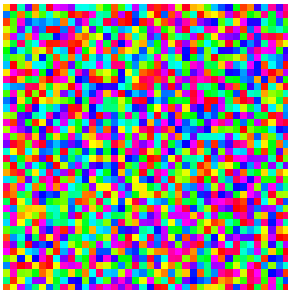
# The algorithm

- ▶ Form  $\bar{K}$ : just copy the columns of  $A_k$
- ▶ Compute  $K$ : rank profile of  $\bar{K}$

# The algorithm

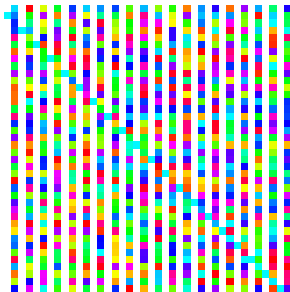
- ▶ Form  $\bar{K}$ : just copy the columns of  $A_k$
- ▶ Compute  $K$ : rank profile of  $\bar{K}$
- ▶ Apply the similarity transformation  $K^{-1}A_kK$

# Example

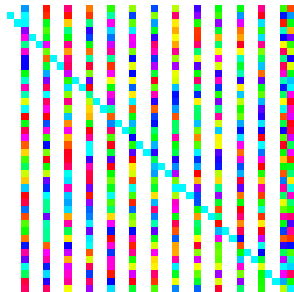




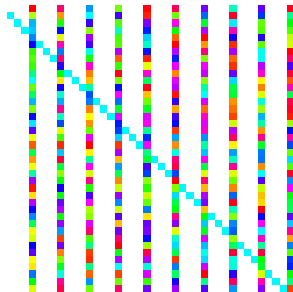
# Example



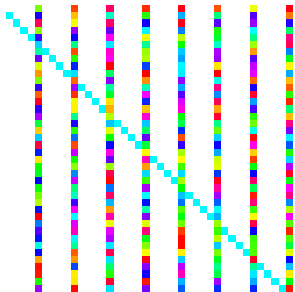
# Example



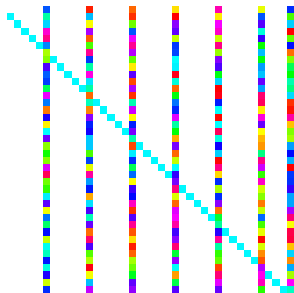
# Example



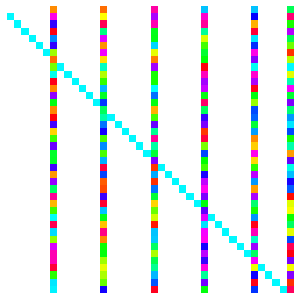
# Example



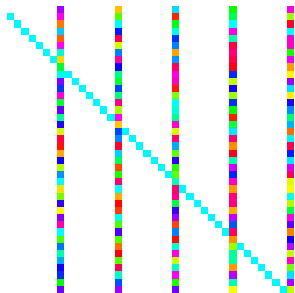
# Example



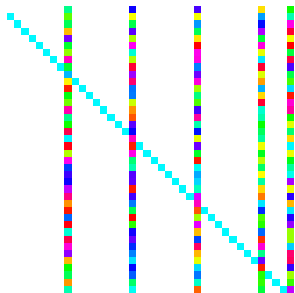
# Example



# Example

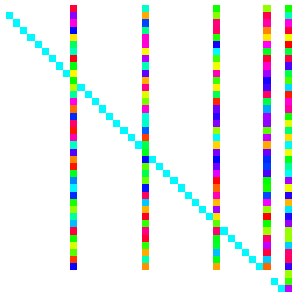


# Example

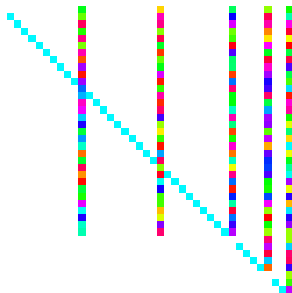




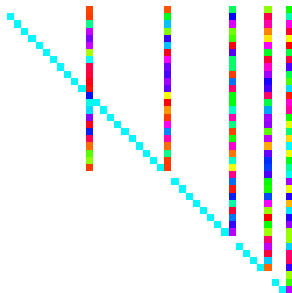
# Example



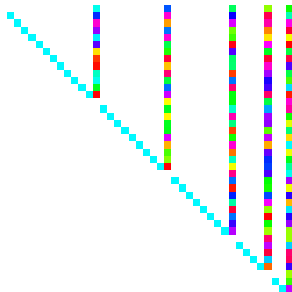
# Example



# Example



# Example



## Lemma

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

## Lemma

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

How to use fast matrix arithmetic ?







# Reduction to Matrix multiplication

Similarity transformation: parenthesing

$$K^{-1}AK = Q'^{-1} \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} P'^{-1}Q \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} PQ' \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} P'$$

# Reduction to Matrix multiplication

Similarity transformation: parenthesing

$$K^{-1}AK = Q'^{-1} \left( \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \left( P'^{-1}Q \left( \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \left( PQ' \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \right) \right) \right) \right) P'$$

# Reduction to Matrix multiplication

Similarity transformation: parenthesing

$$K^{-1}AK = Q'^{-1} \left( \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \left( P'^{-1}Q \left( \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \left( PQ' \begin{bmatrix} I & * \\ 0 & * \end{bmatrix} \right) \right) \right) \right) P'$$

$\Rightarrow \mathcal{O} \left( k \left( \frac{n}{k} \right)^\omega \right)$

# Reduction to Matrix multiplication

Similarity transformation: parenthesing

$$K^{-1}AK = Q'^{-1} \left( \left[ \begin{array}{cc} I & * \\ 0 & * \end{array} \right] \left( P'^{-1} Q \left( \left[ \begin{array}{cc} I & * \\ 0 & * \end{array} \right] \left( P Q' \left[ \begin{array}{cc} I & * \\ 0 & * \end{array} \right] \right) \right) \right) \right) P'$$

$\Rightarrow \mathcal{O} \left( k \left( \frac{n}{k} \right)^\omega \right)$

Overall complexity: summing for each iteration:

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

# A new type of reduction

$$xI_n - A$$

dimension =  $n$   
degree = 1

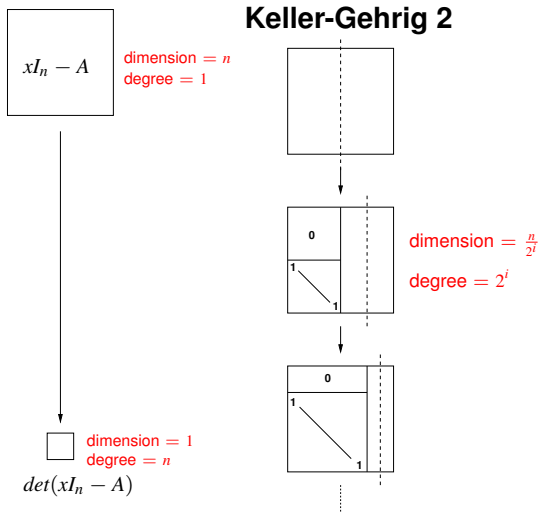


$$\square$$

dimension = 1  
degree =  $n$

$$\det(xI_n - A)$$

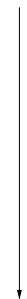
# A new type of reduction



# A new type of reduction

$$xI_n - A$$

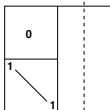
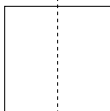
dimension =  $n$   
degree = 1



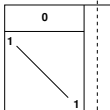
$$\det(xI_n - A)$$

dimension = 1  
degree =  $n$

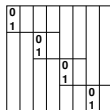
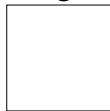
## Keller-Gehrig 2



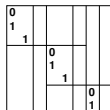
dimension =  $\frac{n}{2^i}$   
degree =  $2^i$



## New algorithm



dimension =  $\frac{n}{k}$   
degree =  $k$



# Conclusion

## Reductions to a building block

**Matrix Mult:** block rec.  $\sum_{i=1}^{\log n} n \left(\frac{n}{2^i}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Gauss, REF)

**Matrix Mult:** Iterative  $\sum_{k=1}^n n \left(\frac{n}{k}\right)^{\omega-1} = \mathcal{O}(n^\omega)$  (Frobenius)

**Linear Sys:** over  $\mathbb{Z}$  (Hermite Normal Form)

## Size/dimension compromises

- ▶ Hermite normal form : rank 1 updates reducing the determinant
- ▶ Frobenius normal form : degree  $k$ , dimension  $n/k$  for  $k = 1 \dots n$