

# From BLAS routines to finite field exact linear algebra solutions



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## Main goals

Solve Linear Algebra problems *exactly* using **BLAS routines**

Implementation in **LinBox** library



- Focus on finite fields
- Use matrix multiplication and BLAS routines as kernel
- Fast exact linear algebra routines (triangular solver, LSP factorization)

## Finite field computations via BLAS routines

Main idea [Dumas-Gautier-Pernet 02]

Convert data from finite field to double

Direct call to BLAS

Convert back the result

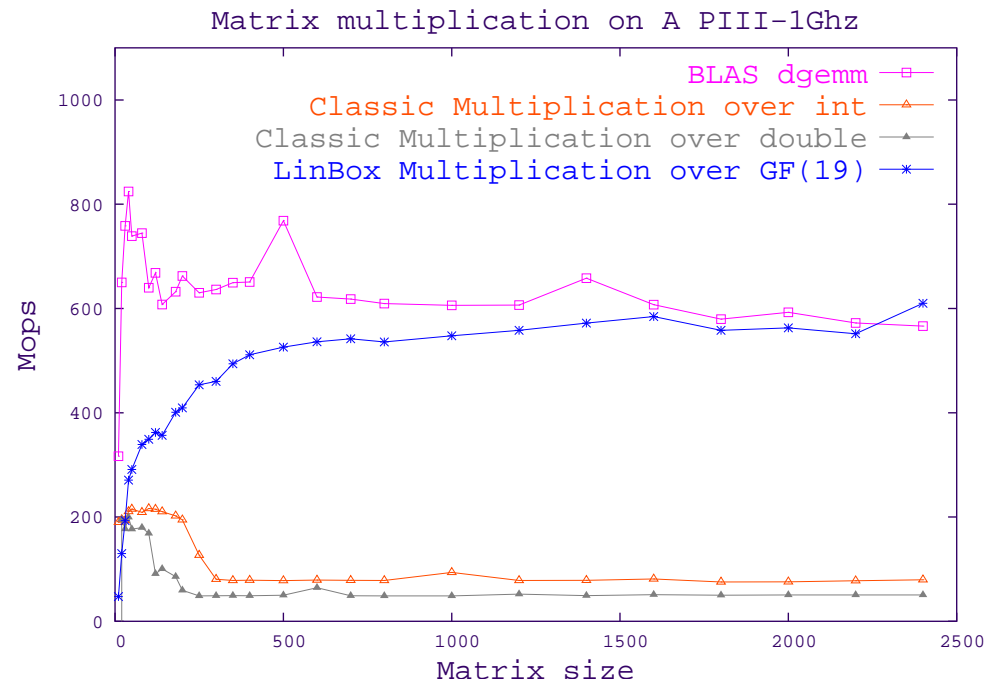
- Only one reduction
- Certify data hold over **53** bits
- Use division-free BLAS routines only

## Illustration with matrix multiplication

- ( $m$  by  $k$  matrix)  $\times$  ( $k$  by  $n$  matrix) over  $\mathbf{GF}(p)$ .

certificate :  $k(p - 1)^2 < 2^{53}$

- Performances with  $m = n = k$  and  $p = 19$  :



- Even better with Strassen-Winograd algorithm [Dumas-Gautier-Pernet 02]

## Our extension to a triangular solver with matrix r.h.s.

### Problem:

- Certificate is  $p^k < 2^{53}$  instead of  $kp^2 < 2^{53}$  for matrix multiplication  
Direct call to BLAS only is too restrictive (only small matrices)

### Solution:

- Use a block recursive algorithm
- Decrease matrix size until certification
- Use BLAS-based matrix multiplication routine to reconstruct the solution

## Block recursive algorithm

- Solve  $\mathbf{AX} = \mathbf{B}$  over  $\mathbf{GF}(p)$
- While no certification

$$\begin{array}{|c|c|} \hline \mathbf{A}_1 & \mathbf{A}_2 \\ \hline & \mathbf{A}_3 \\ \hline \end{array} \times \begin{array}{|c|} \hline \mathbf{X}_1 \\ \hline \mathbf{X}_2 \\ \hline \end{array} = \begin{array}{|c|} \hline \mathbf{B}_1 \\ \hline \mathbf{B}_2 \\ \hline \end{array}$$

$$\left\{ \begin{array}{l} \text{solve } \mathbf{A}_3\mathbf{X}_2 = \mathbf{B}_2 \\ \mathbf{B}_1 \leftarrow \mathbf{B}_1 - \mathbf{A}_2\mathbf{X}_2 \\ \text{solve } \mathbf{A}_1\mathbf{X}_1 = \mathbf{B}_1 \end{array} \right. \quad \begin{array}{l} \text{recursive call} \\ \text{BLAS-based MM} \\ \text{recursive call} \end{array}$$

- Now, how to solve small (certified) linear systems?

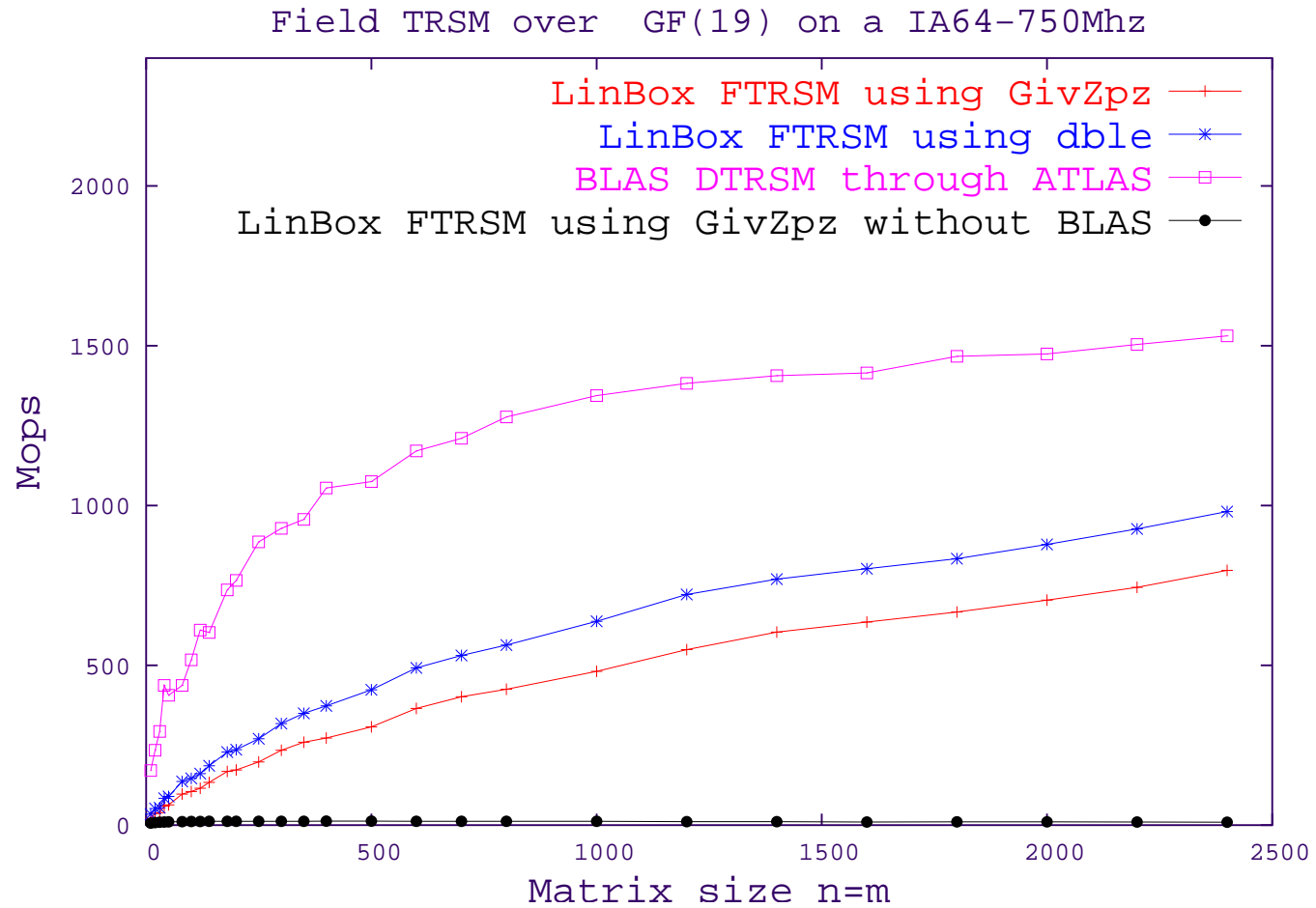
## Solving a certified triangular linear system

- Certified as soon as  $(p - 1)p^{m-1} < 2^{53}$  ( $m$  = row dimension of small system  $\mathbf{AX} = \mathbf{B}$ )
- Let  $\mathbf{A} = \mathbf{UD}$  over  $\mathbf{GF}(p)$   
with  $\mathbf{D}$  a diagonal matrix and  $\mathbf{U}$  a unit upper triangular matrix

Solve  $\mathbf{UY} = \mathbf{B}$  using the *dtrsm* BLAS routine

- Return  $\mathbf{X} = \mathbf{D}^{-1}\mathbf{Y}$  over  $\mathbf{GF}(p)$

## Performances over **GF(19)** on Intel Itanium





In summary, we have just seen

- BLAS-based matrix multiplication
- BLAS-based triangular solver with matrix r.h.s.

Now, let us see

- LSP factorization

## What is LSP factorization?

- **LSP** matrix factorization [Bini-Pan 94]

**L** - lower triangular matrix with 1's on the main diagonal

**S** - reduces to an u.t. matrix with nonzero diagonal entries when zero rows deleted

**P** - permutation matrix

- Exemple over **GF(7)**:

$$\begin{bmatrix} 1 & 3 & 5 & 2 & 4 & 6 \\ 3 & 2 & 1 & 6 & 5 & 4 \\ 3 & 3 & 6 & 0 & 1 & 2 \\ 5 & 3 & 3 & 6 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & & & & & \\ 3 & 1 & & & & \\ 1 & & 1 & & & \\ 1 & & 2 & 1 & & \end{bmatrix} \times \begin{bmatrix} 3 & 1 & 5 & 6 & 2 & 4 \\ & 2 & 1 & 3 & 5 & 4 \\ & & 1 & 1 & 5 & 3 \end{bmatrix} \times \begin{bmatrix} & 1 & & & & \\ 1 & & & & & \\ & & 1 & & & \\ & & & 1 & & \\ & & & & 1 & \\ & & & & & 1 \end{bmatrix}$$

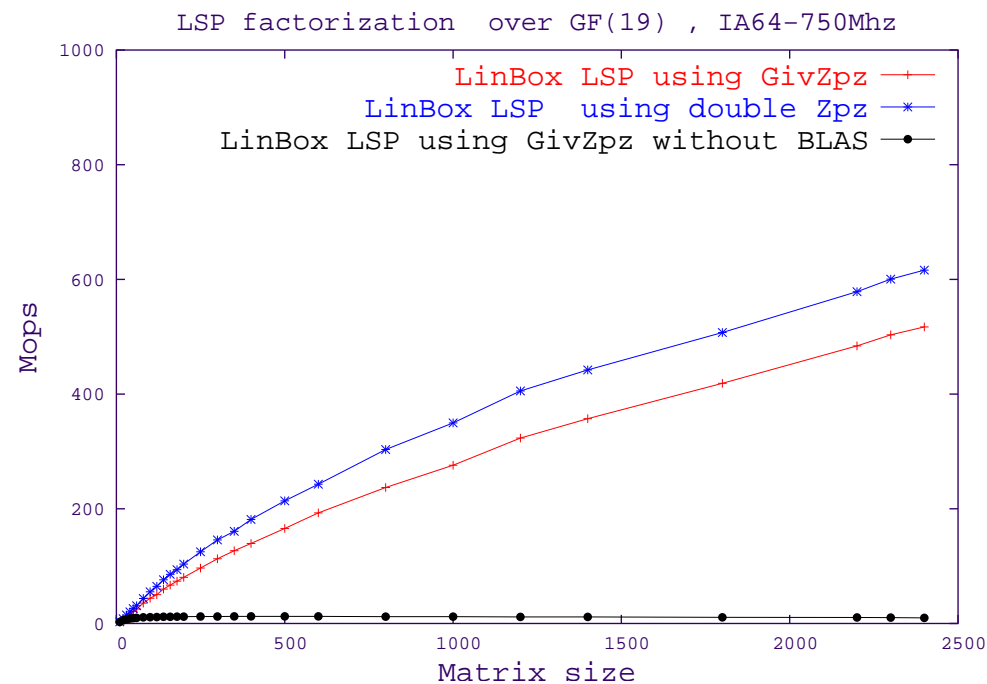
## LSP factorization via matrix multiplication

- Recursive algorithm [Ibarra-Moran-Hui 82]:
- Partition  $\mathbf{A} = \begin{bmatrix} \mathbf{A}_1 \\ \mathbf{A}_2 \end{bmatrix}$  and factor  $\mathbf{A}_1 = \mathbf{L}_1 \mathbf{S}_1 \mathbf{P}_1$
- Partition  $\mathbf{S}_1 = [\mathbf{S}'_1 \ \mathbf{B}]$  and  $\mathbf{A}_2 \mathbf{P}_1^{-1} = [\mathbf{C} \ \mathbf{D}]$
- Solve  $\mathbf{G} \mathbf{S}'_1 = \mathbf{C}$  and factor  $\mathbf{D} - \mathbf{G} \mathbf{B} = \mathbf{L}_2 \mathbf{S}_2 \mathbf{P}_2$
- Reconstruct with formula:

$$\mathbf{A} = \underbrace{\begin{bmatrix} \mathbf{L}_1 & & \\ & \mathbf{G} & \\ & & \mathbf{L}_2 \end{bmatrix}}_{\mathbf{L}} \times \underbrace{\begin{bmatrix} \mathbf{S}'_1 & \mathbf{B} \mathbf{P}_2^{-1} \\ & \mathbf{S}_2 \end{bmatrix}}_{\mathbf{S}} \times \underbrace{\begin{bmatrix} \mathbf{I} & & \\ & \mathbf{P}_2 & \\ & & \mathbf{P}_1 \end{bmatrix}}_{\mathbf{P}}$$

## Our implementation

- Solve  $\mathbf{GS}'_1 = \mathbf{C}$  by using BLAS-based triangular solver
- Compute  $\mathbf{D} - \mathbf{GB}$  by using BLAS-based matrix multiplication
- Performances over  $\mathbf{GF}(19)$  on Intel Itanium:



In summary, we have just seen

- BLAS-based matrix multiplication
- BLAS-based triangular solver with matrix r.h.s.
- BLAS-based LSP factorization

Now, let us see

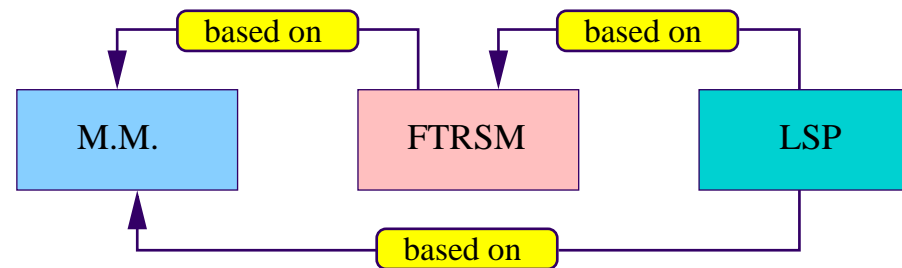
- Availability in LinBox and application to minimal matrix polynomial

## Integration in LinBox library

- FFLAS package [Dumas-Gautier-Pernet 02]
- Fast BLAS-based triangular solver (BLAS-like interface)  
template < class Field >  
void Field\_trsm( const Field& F,  
int m, int n,  
typename Field::Element \* B, int ldb,  
typename Field::Element \* A, int lda,  
typename Field::Element \* X, int ldx,  
Triangular tr,  
Unitary un,  
Side si);
- Fast BLAS-based LSP factorization
- Genericity over the domain  $\mathbf{GF}(\mathbf{p})$  and dense matrices (with BLAS-like storage)

## Features of our LinBox implementation routines

- Easier to implement higher level algorithms based on matrix multiplication
- Speeding up matrix multiplication  $\implies$  faster routines



- Example: computation of minimal matrix polynomial

## Minimal matrix polynomial $\Pi_{\mathbf{A}}(\mathbf{x})$ of a square matrix $\mathbf{A}$

- Important algorithm in LinBox (Krylov/Lanczos approach)

Two main steps of the algorithm:

- Compute the first terms of  $\mathbf{UV}$ ,  $\mathbf{UAV}$ ,  $\mathbf{UA}^2\mathbf{V}$ , ...

- Deduce  $\Pi_{\mathbf{A}}(\mathbf{x})$  by computing a matrix approximant of 
$$\begin{bmatrix} \sum_i (\mathbf{UA}^i\mathbf{V})\mathbf{x}^i & \mathbf{0} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

[Turner's PhD Thesis 02]



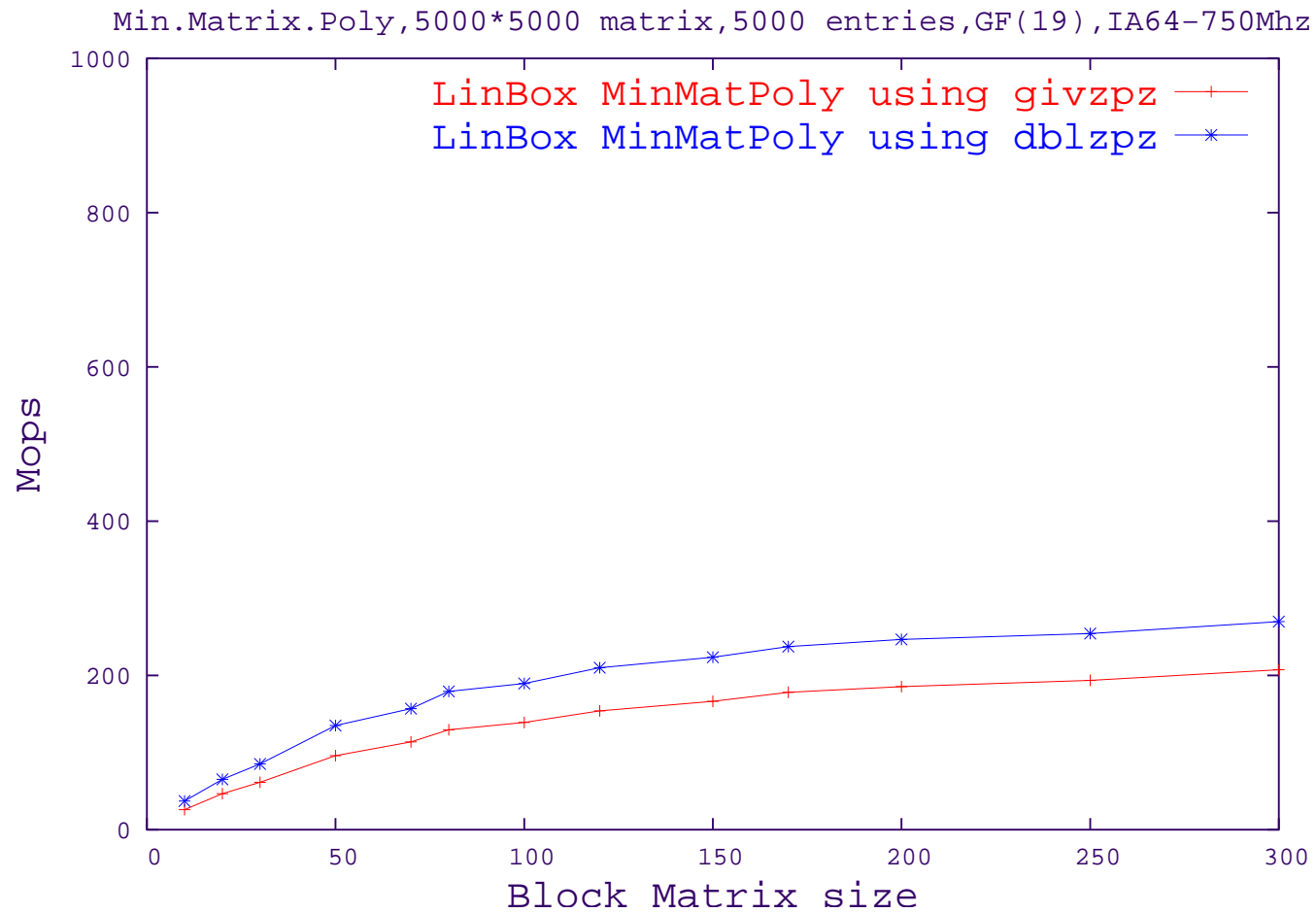
## Matrix approximant algorithm

- Beckermann-Labahn's algorithm via matrix multiplication [Giorgi-Jeannerod-Villard 03]
- Iterative algorithm which computes approximant  $\mathbf{M}(\mathbf{x})$  s.t.

$$\mathbf{M}(\mathbf{x})\mathbf{F}(\mathbf{x}) = \mathbf{O}(\mathbf{x}^\sigma) \text{ in } \sigma \text{ steps}$$

- Main operations involved at step  $\mathbf{k}$ 
  - $\mathbf{k}$  calls to matrix multiplication
  - $\mathbf{1}$  call to **LSP** factorization
  - $\mathbf{k}$  calls to triangular system solving
- We have used our LinBox BLAS-based routines to implement this algorithm

# First performances over **GF(19)** on Intel Itanium



## Conclusion and future work

- Significant improvement for some linear algebra problems over  $\mathbf{GF}(p)$
- Implementation in LinBox library [\[www.linalg.org\]](http://www.linalg.org)
- Extension to sparse matrices
- Extension to other algorithms using matrix multiplication