

# The LinBox Project for Linear Algebra Computation

## A Practical Tutorial

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# Goals for Today

I want to convince you that LinBox is...

- **the** tool for exact linear algebra
- easy to install and use (if you want it to be)
- worth getting involved with

## 1 Foundations

- History
- Major Contributors
- Mathematical Background

## 2 Design

- Middleware Heirarchy
- Genericity

## 3 Practicalities

- Levels of Use
- Details

## 4 Conclusions

## Significant Publications

- 1986 Wiedemann: *Solving sparse linear equations over finite fields*
- 1991 Kaltofen & Saunders: On Wiedemann's Algorithm
- 1993 Coppersmith: Block Lanczos, Block Wiedemann
- 1995 Montgomery: Implementation of Block Lanczos
- 1997 Villard: Analysis of Block Wiedemann
- 2001 Giesbrecht; Dumas, Saunders, & Villard: Smith forms
- 2001 Chen, Eberly, Kaltofen, Saunders, Turner, Villard:  
*Efficient Matrix Preconditioners for Black Box Linear Algebra*
- 2002 Dumas, Gautier, Giesbrecht, Giorgi, Hovinen, Kaltofen,  
Saunders, Turner, Villard:  
*Linbox: A Generic Library For Exact Linear Algebra*
- 2002 Dumas, Gautier, Pernet:  
*FFLAS: Finite Field Linear Algebra Subroutines*

# LinBox Milestones

- 2000 Initial design meetings
- 2002 World Scientific paper
- 2004 BLAS is thoroughly integrated
- 2005 LinBox 1.0 Released
  - JGD gives tutorial at ISSAC in Beijing
  - LinBox included in open-source computer algebra CD
- 2006 Maple interface and web computation server
- 2007 SAGE integration
- 2008 Linbox 1.5

# What is LinBox?

- C++ Library for Exact Computational Linear Algebra
- Open-source (LGPL), international research project
- Generic, using C++ templates
- Middleware — uses fast, low-level libraries, used by higher-level CA systems
- Originally for sparse & structured blackbox computations
- Now incorporates dense as well

# Fathers of LinBox



Dave Saunders



Erich Kaltofen



Gilles Villard



Mark Giesbrecht

# Developers

- Jean-Guillaume Dumas
- Bradford Hovinen
- Will Turner
- David Pritchard
- Clement Pernet
- Pascal Giorgi
- Zhendong Wang
- William Stein
- Mike Abshoff



# Wiedemann's Algorithm

## The Problem

Given nonsingular  $A \in \mathbb{F}^{n \times n}$  and  $\mathbf{v} \in \mathbb{F}^n$  find  $\mathbf{x} \in \mathbb{F}^n$  such that  $A\mathbf{x} = \mathbf{v}$ .

# Wiedemann's Algorithm

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## The Algorithm

First, project on the right:

The sequence

$$\mathbf{v}, \quad A\mathbf{v}, \quad A^2\mathbf{v}, \quad A^3\mathbf{v}, \quad \dots$$

is linearly recurrent of degree  $n$ .

Note:  $A^i\mathbf{v} = A \cdot (A^{i-1}\mathbf{v}) = A \cdot (A \cdot (A \cdots A\mathbf{v}))$

# Wiedemann's Algorithm

## The Problem

Given nonsingular  $A \in \mathbb{F}^{n \times n}$  and  $\mathbf{v} \in \mathbb{F}^n$  find  $\mathbf{x} \in \mathbb{F}^n$  such that  $A\mathbf{x} = \mathbf{v}$ .

## The Algorithm

Then, project on the left **with random  $\mathbf{u} \in \mathbb{F}^n$** :

The sequence

$$\mathbf{u}^T \mathbf{v}, \mathbf{u}^T A \mathbf{v}, \mathbf{u}^T A^2 \mathbf{v}, \mathbf{u}^T A^3 \mathbf{v}, \dots$$

has the same linear recurrence **with high probability**.

# Wiedemann's Algorithm

## The Problem

Given nonsingular  $A \in \mathbb{F}^{n \times n}$  and  $\mathbf{v} \in \mathbb{F}^n$  find  $\mathbf{x} \in \mathbb{F}^n$  such that  $A\mathbf{x} = \mathbf{v}$ .

## The Algorithm

Compute  $f \in \mathbb{F}[x]$ , the minimum polynomial of  $(\mathbf{u}^T A^i \mathbf{v})_{i \geq 0}$ .

$$\begin{aligned} f(x) &= -1 + f_1 x + f_2 x^2 + \cdots + f_n x^n \\ 0 &= -\mathbf{u}^T \mathbf{v} + f_1 \mathbf{u}^T A \mathbf{v} + f_2 \mathbf{u}^T A^2 \mathbf{v} + \cdots + f_n \mathbf{u}^T A^n \mathbf{v} \end{aligned}$$

# Wiedemann's Algorithm

## The Problem

Given nonsingular  $A \in \mathbb{F}^{n \times n}$  and  $\mathbf{v} \in \mathbb{F}^n$  find  $\mathbf{x} \in \mathbb{F}^n$  such that  $A\mathbf{x} = \mathbf{v}$ .

## The Algorithm

W.h.p.  $f$  is the minimum polynomial of  $(A^i \mathbf{v})_{i \geq 0}$ .

$$\begin{aligned} f(x) &= -1 + f_1 x + f_2 x^2 + \cdots + f_n x^n \\ 0 &= -\mathbf{u}^T \mathbf{v} + f_1 \mathbf{u}^T A \mathbf{v} + f_2 \mathbf{u}^T A^2 \mathbf{v} + \cdots + f_n \mathbf{u}^T A^n \mathbf{v} \\ \mathbf{0} &= -\mathbf{v} + f_1 A \mathbf{v} + f_2 A^2 \mathbf{v} + \cdots + f_n A^n \mathbf{v} \end{aligned}$$

# Wiedemann's Algorithm

## The Problem

Given nonsingular  $A \in \mathbb{F}^{n \times n}$  and  $\mathbf{v} \in \mathbb{F}^n$  find  $\mathbf{x} \in \mathbb{F}^n$  such that  $A\mathbf{x} = \mathbf{v}$ .

## The Algorithm

Rearrange and multiply by  $A^{-1}$ :

$$\begin{aligned}
 f(x) &= -1 + f_1x + f_2x^2 + \cdots + f_nx^n \\
 0 &= -\mathbf{u}^T \mathbf{v} + f_1 \mathbf{u}^T A \mathbf{v} + f_2 \mathbf{u}^T A^2 \mathbf{v} + \cdots + f_n \mathbf{u}^T A^n \mathbf{v} \\
 \mathbf{0} &= -\mathbf{v} + f_1 A \mathbf{v} + f_2 A^2 \mathbf{v} + \cdots + f_n A^n \mathbf{v} \\
 \mathbf{x} = A^{-1} \mathbf{v} &= f_1 \mathbf{v} + f_2 A \mathbf{v} + \cdots + f_n A^{n-1} \mathbf{v}
 \end{aligned}$$

# Black Box Approach



## “Usual” Goals

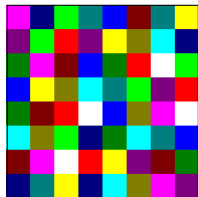
- $O(n)$  black box calls
- $O(n^2)$  other work
- $O(n)$  space

## Good Black Box Algorithms for:

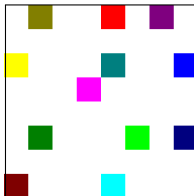
minimum polynomial, rank, determinant, linear system solving, Smith normal form, ...

# Black Box Classes

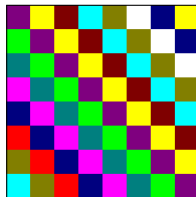
Common costs for black box vector apply:



Dense  
 $O(n^2)$



Sparse  
 $s$  entries  
 $O(s)$



Toeplitz  
 $O(M(n))$

**Other blackboxes:** Hankel, Sylvester, Diagonal, Hilbert, . . .



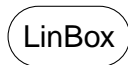
## Block Methods

Similar to before, vectors now replaced by rectangular matrices.

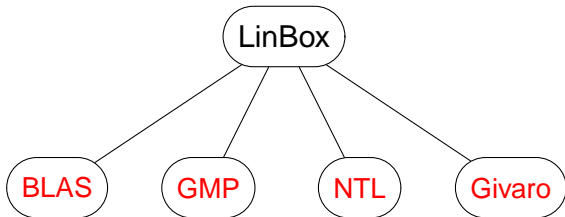


- **Many** mathematical complications
- Allows use of fast dense methods (i.e. BLAS)
- Usually a tradeoff

# LinBox as Middleware

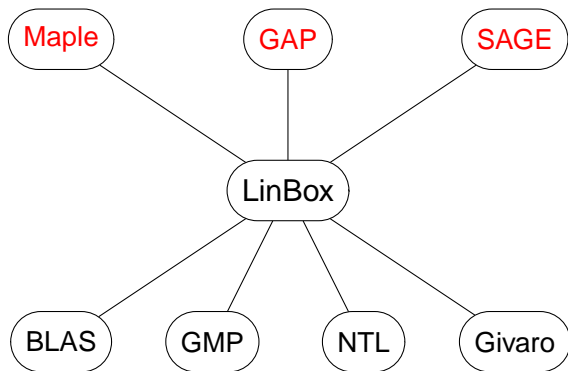


# LinBox as Middleware



## Packages used by LinBox

## LinBox as Middleware



## Higher-level Systems using LinBox

# Basic Linear Algebra Subprograms

Extremely efficient low-level routines for floating-point linear algebra:

## BLAS Levels

- 1 Scalar product, vector-scalar product, vector-vector sum
- 2 Dot product, matrix-vector product
- 3 Matrix-matrix product

Use either:

- Proprietary BLAS — for most architectures (e.g. Intel's MKL)
- Free BLAS (e.g. ATLAS, GotoBLAS)

# GNU Multiple Precision

Provides routines for exact, multiple-precision integer arithmetic.  
Classical, Karatsuba, Toom-Cook, and FFT multiplication  
(with crossovers).

Advantages:

- Many inner loops written in assembly
- Large user community

## NTL, Givaro, ...

- Libraries for number theory (field, ring, polynomial arithmetic).
- Used to work with matrices over finite fields, polynomials, etc.
- **Also** used for structured matrix computations (e.g. Toeplitz)
- Free, open-source, efficient

# Three Choices to Make

## 1 Underlying Domain (“field”)

- Integers, Rationals
- Prime and Prime Power Fields
- Polynomial Rings

## 2 Matrix Representation

- Sparse Black Box
- Dense Explicit
- Structured Black Box

## 3 Algorithm



## Vertical Structure Approach

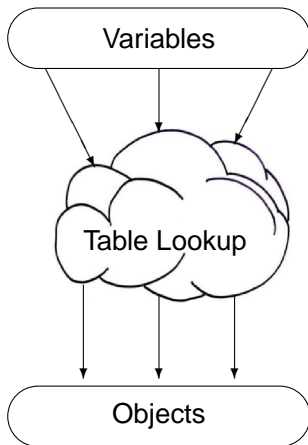
Used e.g. by NTL:

zz_p	GF2	ZZ_pE
zz_pX	GF2X	ZZ_pEX
vec_zz_p	vec_GF2	vec_ZZ_pE
mat_zz_p	mat_GF2	mat_ZZ_pE
⋮	⋮	⋮

- Only one choice given underlying domain!

# Dynamic Type-Checking

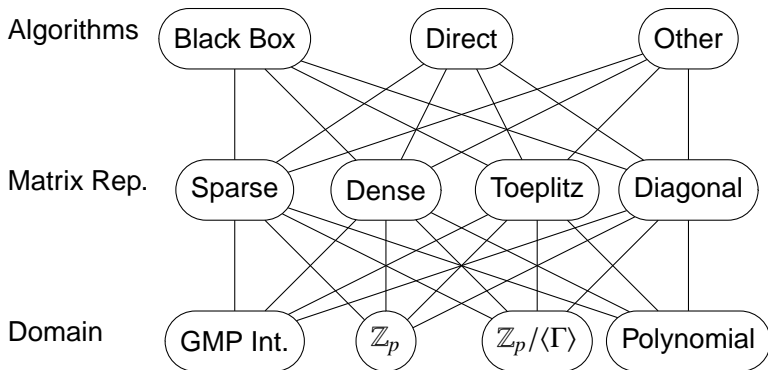
- Traditional polymorphism, required in Java, possible in C++



Have to do a table lookup  
for every variable reference!

## Our Approach: C++ Templates

- All type decisions made at **compile-time**
- Every combination is possible
- Huge efficiency gains over polymorphism (pipelining, inlining)



# Drawbacks to C++ Templates

- Programmers can make **bad decisions**  
(e.g. using elimination algorithm with sparse blackbox matrix)
  
- Binary code bloat  
Every template combination which is used must be compiled.  
Creating a complete, precompiled library is impractical.

# Levels of Use

- 1 Online Computation Server
  - Does not require any installation
- 2 Maple, SAGE, examples directory
  - User sees no C++ code
- 3 solutions directory
  - User chooses field, blackbox, writes in C++
- 4 Expert user
  - Direct programming — complete control

# Installation

Installing LinBox is not difficult!

- Installing ATLAS can be tricky
- All LinBox dependencies can be found in Debian repo.
- Getting maximal efficiency is inherently difficult
- Middleware is inherently troublesome

# Folder Structure

examples	Ready-to-go programs for typical problems
interfaces	Connections to Maple, SAGE, etc.
linbox	Library code
/field	Underlying domains
/blackbox	Matrix representations
/algorithms	Heart of the library
/solutions	Code for typical problems
/randiter	Random element generation
/util	General utilities (e.g. I/O)
doc	Documentation generation
tests	Correctness checks and benchmarks

# Input/Output

If using LinBox at “Level” 1 or 2, I/O is handled automatically.

Otherwise, use the `MatrixStream` class.

- Automatic recognition of many formats
- Outputs (row,column,value) “triples” or a single dense array
- Connects to most matrix representations
- Allows any rep. to be read from any file

Similar ideas for output have been proposed; not yet implemented.



# What LinBox is Great At

LinBox is great at computing

{rank, determinant, linear solution, characteristic polynomial, Smith form}

for matrices which are

{dense, sparse, structured}

over

{integers, rationals, prime fields, extension fields, polynomials}.

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## What LinBox is *not* Great At

- Inverse computation
- Nullspace computation
- Support for GF(2)
- Parallelism
- Block methods

# Software Engineering

LinBox has a rapidly growing user base (!)

**We need some software engineering!**

- Configure/Install (autohell)
- Removing legacy code
- Restructuring matrix/blackbox distinction
- Interfaces

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**ANY VOLUNTEERS?**