

Synthesizing Numerical Linear Algebra using Julia

julia

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Objective

Improve efficiency of linear algebra routines:

- LAPACK / BLAS: most widely recognized linear algebra library
 - However, provides different routine for each data type and precision
- Each hardware provides own package

Provide agonistic implementations of LAPACK / BLAS routines:

- Recent efforts in C++ and Python
 - Ex: C++26 has all BLAS operations in language standard [3]

Use Julia Language for composability:

- Multiple-dispatch and type inference
 - LLVM dynamically generates optimized code for different data / hardware types [2]
- Provide single API
 - Ease of usage benefits
 - Shorter development time
- Extendibility
 - Update to new hardware
 - Extend to new data types

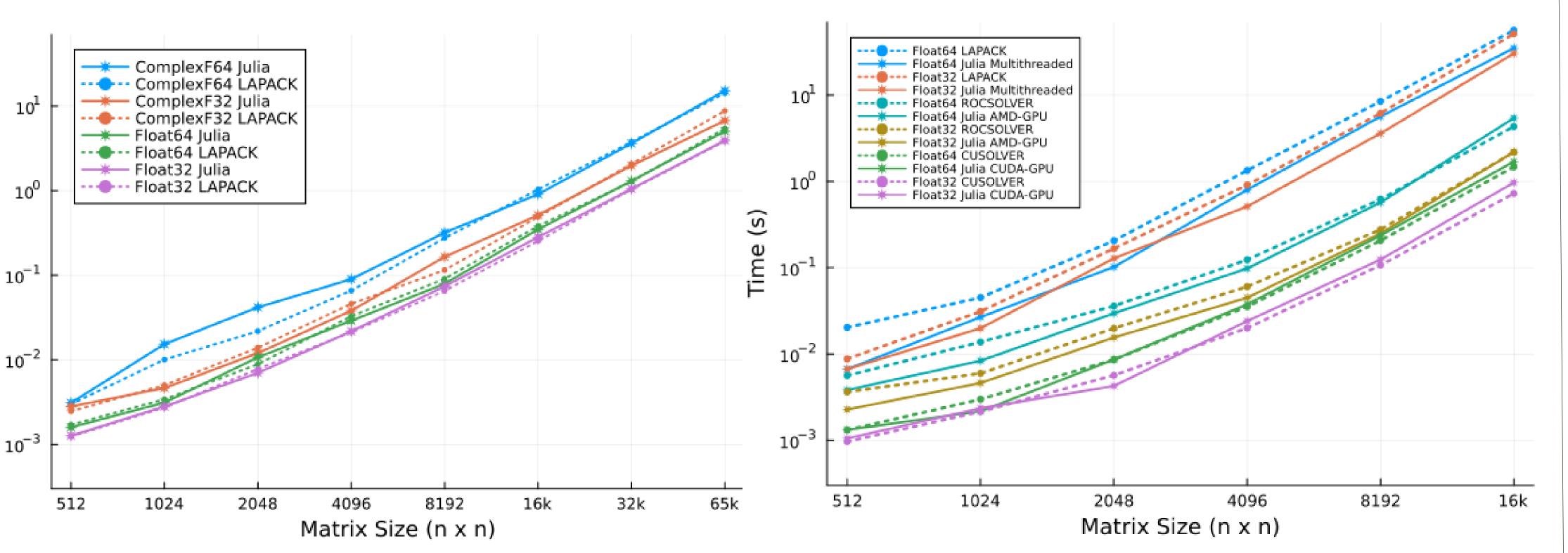
References

- [1] Bezanson, Julia: Dynamism and Performance. *Proc. ACM Program. Lang.*, oct 2018.
- [2] Bezanson, Array Operators Using Multiple Dispatch: In Proceedings of ACM SIGPLAN International Workshop on Libraries, Languages, and Compilers for Array Programming
 [3] cppreference. C++26: Basic linear algebra algorithm, 2024
 [4] Danisch, GPUArrays

A generic API for linear algebra in Julia provides composability without sacrificing performance

Generic **larfb** function matches performance of data-type specific LAPACK implementation

Single-API **unmqr** closely matches performance of both LAPACK (CPU) and CUSOLVER/ROCSOLVER (GPU)



Performance across data types

(Intel Ice Lake Processor)

Performance across Hardware

(Intel CascadeLake CPU and V100/MI100 GPU)

What's next?

- 1. Expand library to include more routines -- LU Factorization, QR factorization, etc.
 - 2. Add support to mixed precisions computation
- 3. Further optimizations for parallel computing -- KernelAbstractions, Dagger, etc.

Methods

Implemented larfb and unmqr functions in Julia

- **unmqr**: applies orthogonal matrix Q of a QR factorization to generic rectangular matrix A:
- $\bullet \qquad A + AYTY^T = AQ$

- ullet Y is block-householder factorization forming Q
- T is scalar factors of elementary reflectors
- larfb: performs the individual projections; used in unmqr
- Publicly available in DLA.jl

Abstract Array interface allows for Unified API [4]

- Supports arbitrary data types
 - Precision (8/16/32/64 bit)
 - Data type (Integer/ Float/ Complex)
- Dispatch to CPU / GPU with minimal changes to API
 - Separate implementations historically

| Vendor and Family | Intel Ice Lake | Intel CascadeLake | AMD Milan |
|----------------------|----------------|----------------------|--------------|
| Model | 6330 | 6248 | 7713 |
| Sockets(s) | 2 | 2 | 2 |
| Cores per Socket | 28 | 20 | 64 |
| Clock Speed | 2 GHz | 2.5 GHz | 2.0GHz |
| DDR Memory Size | 1 TB | 384 GB | 256GB |
| L3 Cache Size | 84 MiB | 27.5 MiB | 512 MiB |
| GPU type | NVIDIA A100 | NVIDIA V100 | AMD MI100 |

Hardware Specifications