

An introduction to numerical methods with BLAS

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Outline

- Setting up a system to compile a BLAS library and link it with executables
- Data representation for vector and matrices
- The interface of BLAS
- Effects of caches and cache aware programming
- Resources for testing and benchmarking





Not all evaluations perform the same!

Consider a simple matrix-vector multiplication:

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix}. \tag{1}$$

The evaluation can be reduced with either 2 dot product or 2 scalar-matrix product operations:

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} \begin{pmatrix} 1 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} \\ \begin{pmatrix} 3 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} \end{pmatrix}, \qquad \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 1 \begin{pmatrix} 1 \\ 3 \end{pmatrix} + 2 \begin{pmatrix} 2 \\ 4 \end{pmatrix}.$$

Are there differences in the efficiency of the 2 evaluation methods?





Setting up your system to compile a BLAS library and link it with executables



Preparing your system to compile BLAS

1. Install git





Preparing your system to compile BLAS

- 1. Install git
- 2. Install build tools:
 - o gcc
 - gfortran
 - CMake
 - Ninja





Compiling Netlib BLAS

Fetching the code

• Clone the git repository from the UL HPC GitLab server.

\$ git clone ssh://git@gitlab.uni.lu:8022/hlst/seminars/blas/lapack.git ~/Documents/blas/lapack

Checkout the branch with the setup for the tutorial

\$ cd ~/Documents/blas/lapack \$ git checkout blas-tutorial



Compiling Netlib BLAS

Build the BLAS components of Netlib LAPACK

Configure the build system with the preset options

\$ cmake --preset default-config

• The command will create a directory 'build' with the instructions to build the software

```
$ cmake --build build/Release --target cblas
$ cmake --build build/Release --target CBLAS/install
```

- The commands will:
 - o build BLAS and CBLAS components
 - o install the libraries in '~/.local/netlib'
- See the 'CMakePresets.json' for more information





Fetch the example code

\$ git clone ssh://git@gitlab.uni.lu:8022/hlst/seminars/blas/blas-tutorial.git ~/Documents/blas/blas-tutorial

• Configure the examples

```
$ cd ~/Documents/blas/blas-tutorial $ cmake --preset default-config
```



Build the example code

\$ cmake --build build/Release --target 00_example_ddot --verbose

• The output is extensive, but we can reproduce the result with the following commands:

```
$ mkdir lib bin
$ gcc -l./include -isystem ${HOME}/.local/netlib/include -O3 -DNDEBUG -o src/00_example_ddot.c.o -c
src/00_example_ddot.c
$ gcc -O3 -DNDEBUG -o src/utils.c.o -c src/utils.c
$ ar qc lib/libutils.a src/utils.c.o && ranlib lib/libutils.a
$ gcc -O3 -DNDEBUG -Wl,--no-as-needed src/00_example_ddot.c.o -o bin/00_example_ddot
-Wl,-rpath,${HOME}/.local/netlib/lib/libutils.a ${HOME}/.local/netlib/lib/libblas.so
${HOME}/.local/netlib/lib/libblas.so -lm
```



Let's break down the compilation commands

\$ gcc -I./include -isystem \${HOME}/.local/netlib/include -O3 -DNDEBUG -o src/00_example_ddot.c.o -c src/00_example_ddot.c

- Create an object file from 'src/00_example_ddot.c'
- The file 'src/00_example_ddot.c' calls function from:
 - o 'utils/h': add the location of the header file in the search path with '-l'
 - o 'cbals.h': add the location of the header file in the system search path with '-isystem'
- Directories included with '-I' are searched before directories included with '-isystem'
- Headers in directories included with '-isystem' must use '<...>' notation



Let's break down the compilation commands

\$ gcc -O3 -DNDEBUG -o src/utils.c.o -c src/utils.c

Creates an object file for the utility library from 'src/utils.c'

\$ ar qc lib/libutils.a src/utils.c.o && ranlib lib/libutils.a

- Composes the object file into a static library
 - Effectively an object file with a lookup table to easily locate functions within the library



Let's break down the compilation commands

 $\$ gcc -O3 -DNDEBUG -WI,--no-as-needed src/00_example_ddot.c.o -o bin/00_example_ddot -WI,-rpath, $\$ {HOME}/.local/netlib/lib/libblas.so -lm

- The command links the executable 'bin/00_example_ddot'
 - -WI,-rpath: Adds the location of the BLAS and CBLAS library in the executable RUNPATH
 - **-WI,--no-as-needed:** ensures that all dependencies are loaded from the correct location
 - o -lm: links with a mathematical library (libm) used in 'utilities.c'
- Dynamic libraries 'libblas.so' for BLAS and 'libcblas.so' and CBLAS will be required in runtime
- Static library 'libutils.a' is linked statically and will not be required in runtime



Run the executable

• Compile the executable with the cmake build scripts:

\$ cmake --build build/Release --target 00_example_ddot --verbose

• After the compilation you will be able to run the resulting executable with:

```
$ ./build/Release/bin/00_example_ddot # print help $ ./build/Release/bin/00_example_ddot 100 2 # An example run
```



Practical session

- Try compiling the CBLAS library
- Try compiling the executable '00_example_ddot'



Software libraries

- BLAS is a typical software library
- Libraries can be used in 2 forms:
 - Static
 - Dynamic

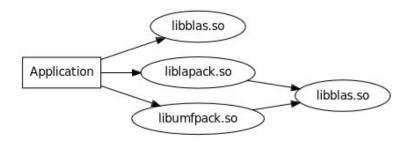
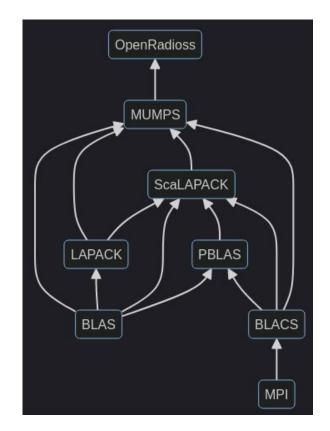
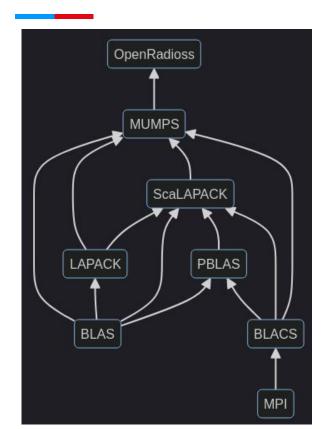


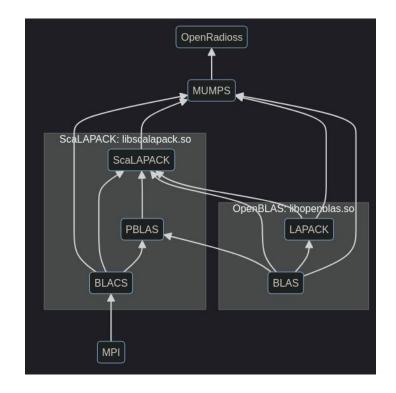
Figure 1: Shared library dependencies of an example application.





Software libraries

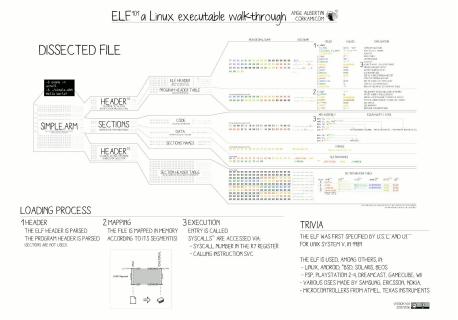








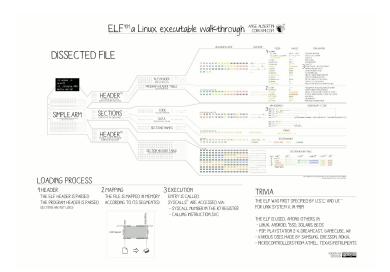
Does a particular 'libopenblas.so' instance implement the CBLAS interface?





To investigate the shared object:

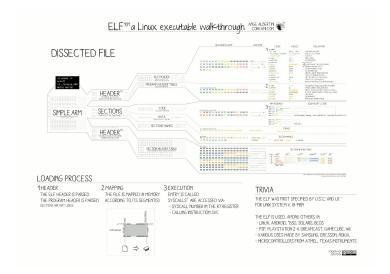
- readelf: display information about ELF files
 - --all: all sections
 - --file-header: information about interoperability
 - --dynamic: dynamically linked libraries and other information
- objdump: display information about objects
 - --syms: information for symbols (functions and variables)
 - --demangle: restore human readable names for objects generated from C++
- nm: list symbols
 - --dynamic: list only export symbols (only for dynamic libraries)





Even extract information about function signatures (needs debug info, -g):

- Read debug info with readelf
 - O --debug-dump=info
- Partially disassemble with objdump
 - o --disassemble
 - o --disassemble-all





Practical session

- Can you break the linking? Try removing the linker option: --no-as-needed

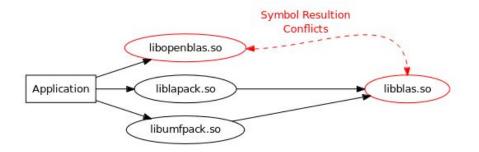


Figure 2: Wrong symbol resolution after relinking the example application.



Data representation for vector and matrices



Matrices

- Computer memory is linear
- Matrices must be linearized:



structure matrix(T)

data: T*

m: integer

n: integer

end structure

$$\begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix} \longrightarrow 1 2 3 4 5 6$$

• How useful is this representation?



Example: Gaussian elimination algorithm

$$A^{(0)} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix} \sim A^{(1)} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 2 \end{pmatrix}$$



Example: Gaussian elimination algorithm

structure matrix(T)

data: T*

ld: integer

m: integer

n: integer

end structure

$$A^{(1)} = \left(\begin{array}{ccc} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 2 \end{array}\right)$$



The interface of BLAS



- Operations organized by computational complexity
 - Level 1: O(n)
 - Level 2: O(n^2)
 - Level 3: O(n^3)
- BLAS supports various number types and numerical precision (first part of function names):
 - single precision (S) with 32-bits,
 - double precision (**D**) with 64-bits,
 - single precision complex (C) with 64-bits, and
 - double precision complex (**Z**) with 128-bits.





 Matrix properties are exploited to save space and reduce memory accesses:

GE:
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \longrightarrow \begin{bmatrix} a_{11} & a_{21} & a_{31} & a_{12} & a_{22} & a_{32} & a_{13} & a_{23} & a_{33} \end{bmatrix}$$

SY:
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{12} & a_{22} & a_{23} \\ a_{13} & a_{23} & a_{33} \end{pmatrix} \longrightarrow \begin{bmatrix} a_{11} & a_{21} & a_{31} & a_{12} & a_{22} & * & a_{13} & * & * \end{bmatrix}$$

TR:
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{pmatrix} \longrightarrow \begin{bmatrix} a_{11} & a_{21} & a_{31} & a_{12} & a_{22} & * & a_{13} & * & * \end{bmatrix}$$

SP:
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{12} & a_{22} & a_{23} \\ a_{13} & a_{23} & a_{33} \end{pmatrix} \longrightarrow \begin{bmatrix} a_{11} & a_{21} & a_{31} & a_{12} & a_{22} & a_{13} \end{bmatrix}$$

TP:
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{pmatrix} \longrightarrow \begin{bmatrix} a_{11} & a_{21} & a_{31} & a_{12} & a_{22} & a_{13} \end{bmatrix}$$



• The type of the matrix representation used forms the 2nd part of the name:

	Storage type		
Algebraic properties	Standard (-)	Banded (B)	Packed (P)
General (G)	GE	GB	
Symmetric (S)	SY	SB	SP
Hermitian (H)	HE	HB	HP
Triangular (T)	TR	TB	TP



- Last part is the type of the operants:
 - V: vector
 - M: matrix
- For instance:

DGEMV:

- D: double precision
- GE: general matrix
- o MV: matrix-vector multiplication

DGEMM:

- o D: double precision
- o GE: general matrix
- MM: matrix-matrix multiplication
- The convention does not work always, especially for Level 1 operations
 - DAXPY: y ax+y





Data representation & BLAS interface

Practical session

- Try exercises 01-03.
- You will need to find and call the appropriate BLAS functions.

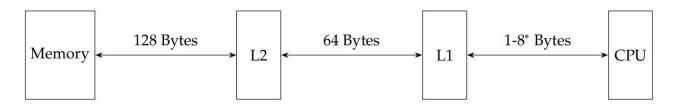


Effects of caches and cache aware programming



- Direct linearization of matrices can degrade performance!
- Caches affect the speed of memory access

```
for ( int i = 0; i < n; ++i ) {
   a[i] = 0;
}</pre>
```



*up to 64 for some special SIMD instructions sets such as AVX-512





- Direct linearization of matrices can degrade performance!
- Caches affect the speed of memory access

```
for ( int i = 0; i < n; ++i ) {
   a[i] = 0;
}</pre>
```

Vectorizable:



- Direct linearization of matrices can degrade performance!
- Caches affect the speed of memory access

```
#pragma omp simd aligned(a:32)
for ( int i = 0; i < 4*n; i+=1 ) {
    a[i] = 0;
}</pre>
Non-vectorizable:
```



- Direct linearization of matrices can degrade performance!
- Caches affect the speed of memory access

```
for ( int i = 0; i < n; i+=1 ) {
    a[4*i] = 0;
    a[4*i+1] = 0;
    a[4*i+2] = 0;
    Non-vectorizable:
    a[4*i+3] = 0;
```



Practical session

- Try exercises 04-05
- Demonstration of the use of 'aligned_alloc' in 'src/read_and_execute.c'
- Hint: for exercise 05 use the function 'get_ld' of the utilities library to get the leading dimension of arrays



Resources for testing and benchmarking



Resources for testing and benchmarking

Resources for BLAS

- Official BLAS webpage: https://www.netlib.org/blas/
- Quick reference (function list): https://www.netlib.org/blas/
- Reference BLAS implementation:
 https://www.netlib.org/lapack/explore-html/d1/df9/group_blas.html

Matrix benchmarks

- Matrix market: https://math.nist.gov/MatrixMarket/
- Matrix market exchange format reader: https://gitlab.uni.lu/hlst/seminars/blas/matrix-market-exchange-formats



Thank you!

Any questions?