



# An introduction to numerical methods with BLAS

Georgios A. Kafanas

High Performance  
Computing &  
Big Data Services



UNIVERSITÉ DU  
LUXEMBOURG

# 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:

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix}. \quad (1)$$

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

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

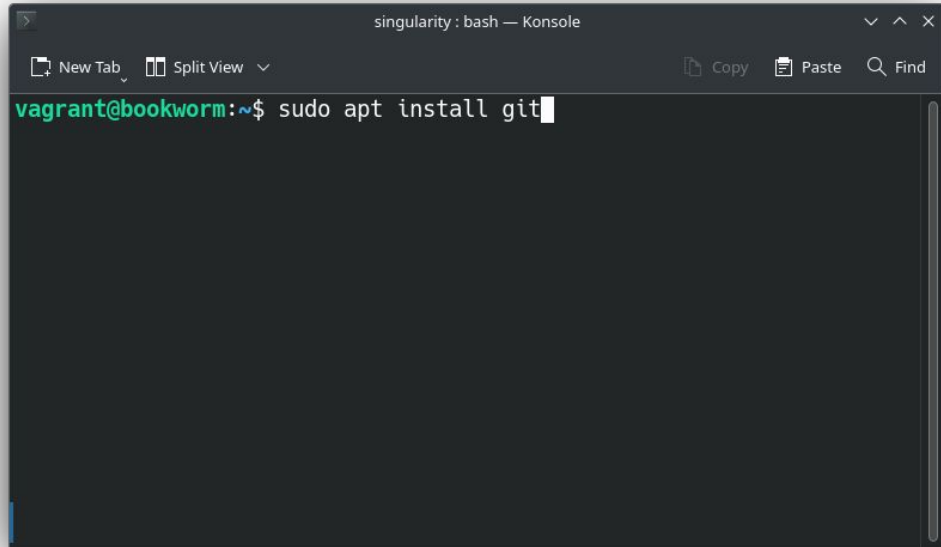
Are there differences in the efficiency of the 2 evaluation methods?



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

# Preparing your system to compile BLAS

## 1. Install git

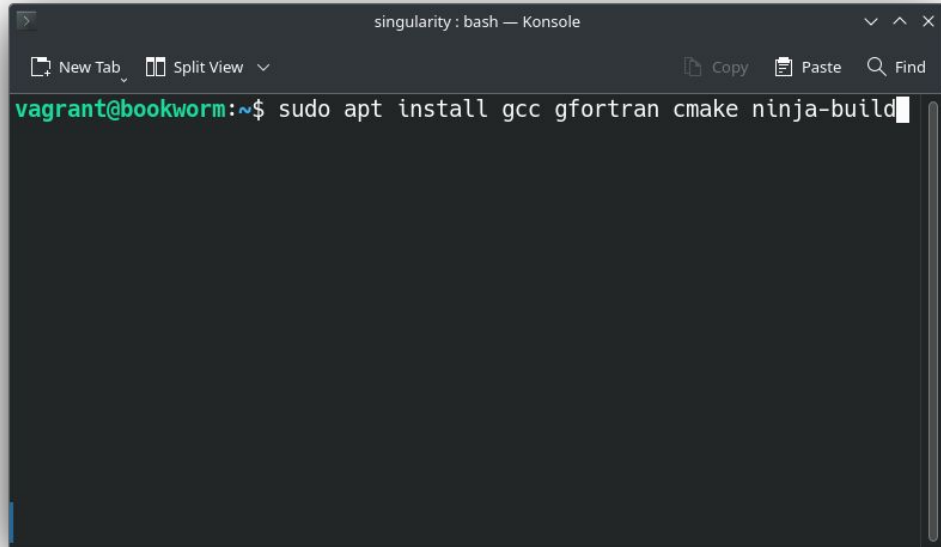


```
singularity : bash — Konsole  
New Tab Split View Copy Paste Find  
vagrant@bookworm:~$ sudo apt install git
```

# Preparing your system to compile BLAS

---

1. Install git
2. Install build tools:
  - gcc
  - gfortran
  - CMake
  - Ninja (optional)



```
singularity : bash — Konsole  
New Tab Split View Copy Paste Find  
vagrant@bookworm:~$ sudo apt install gcc gfortran cmake ninja-build
```

# 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-devel
```

# 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 all -- -j  
$ cmake --build build/Release --target install
```

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



# Compiling the Matrix Market parser library

---

## Build the Matrix Market Exchange Format parser

- 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 all -- -j  
$ cmake --build build/Release --target install
```

- The commands will:
  - build 'matrixmarket' library
  - install the library and its headers in '~/.local/matrix-market'
- See the 'CMakePresets.json' for more information

# Linking with Netlib BLAS

---

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
```

# Linking with Netlib BLAS

## 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 -I./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 lib/libutils.a ${HOME}/local/netlib/lib/libcblas.so  
${HOME}/local/netlib/lib/libblas.so -lm
```

# Linking with Netlib BLAS

---

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:
  - 'utils/h': add the location of the header file in the search path with '-I'
  - '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

# Linking with Netlib BLAS

---

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

# Linking with Netlib BLAS

Let's break down the compilation commands

```
$ gcc -O3 -DNDEBUG -Wl,--no-as-needed src/00_example_ddot.c.o -o bin/00_example_ddot  
-Wl,-rpath,${HOME}/local/netlib/lib lib/libutils.a ${HOME}/local/netlib/lib/libcblas.so  
${HOME}/local/netlib/lib/libblas.so -lm
```

- The command links the executable 'bin/00\_example\_ddot'
  - **-Wl,-rpath:** Adds the location of the BLAS and CBLAS library in the executable RUNPATH
  - **-Wl,--no-as-needed:** ensures that all dependencies are loaded from the correct location
  - **-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

# Linking with Netlib BLAS

---

## 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
```

# Linking with Netlib BLAS

## 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

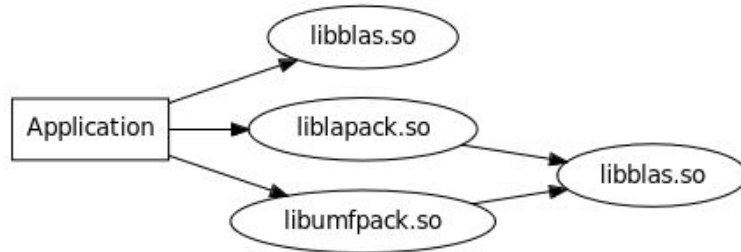
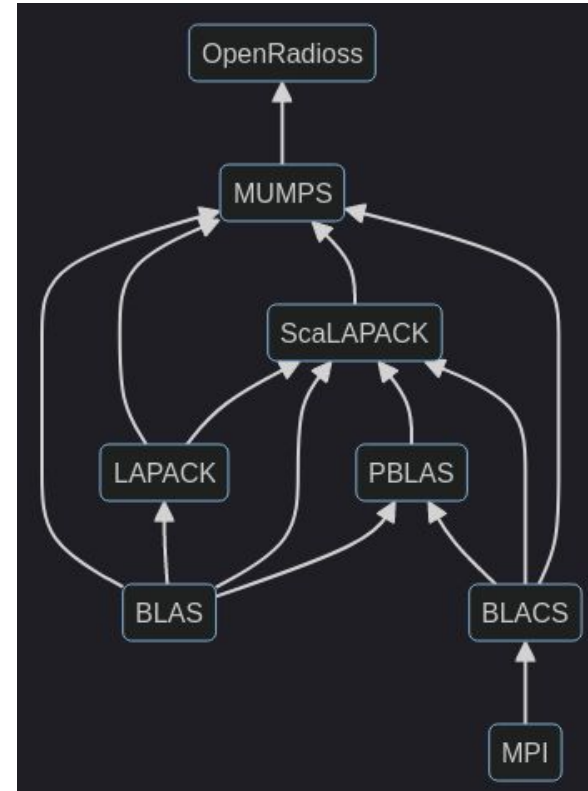
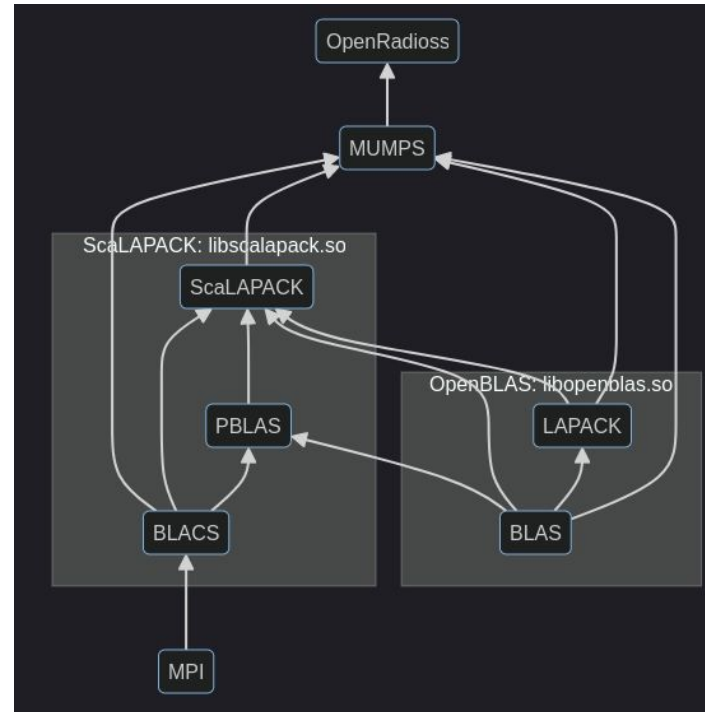
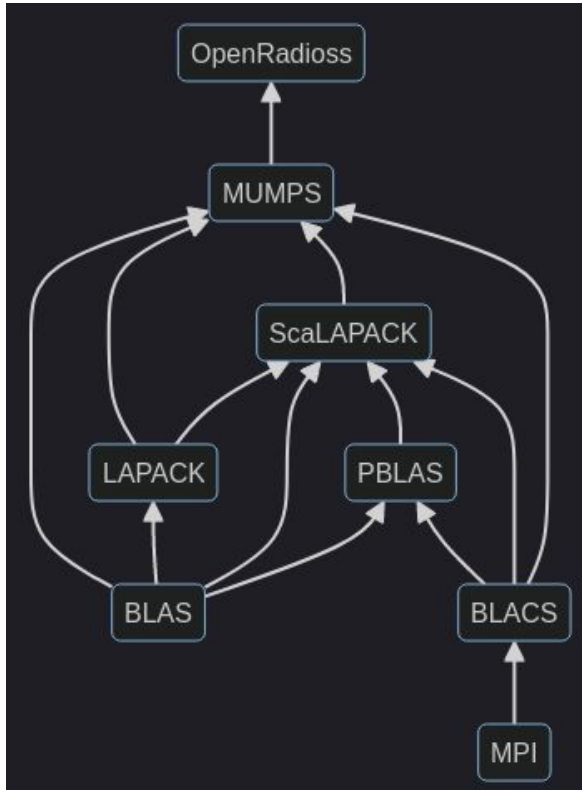


Figure 1: Shared library dependencies of an example application.

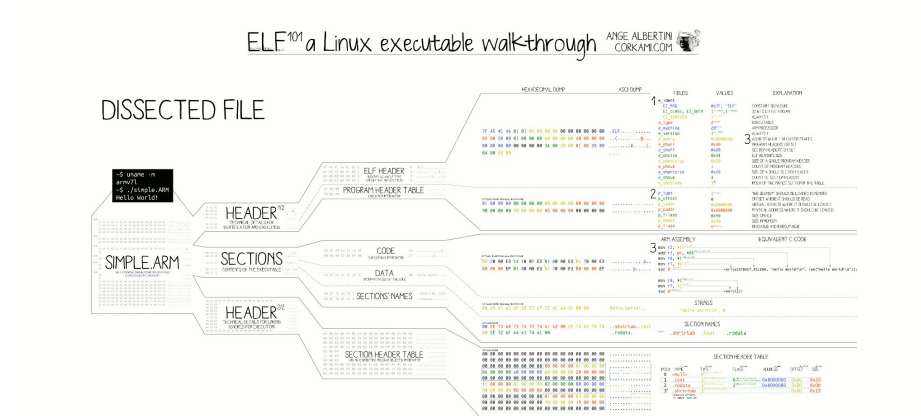


# Software libraries



# Tools for inspecting libraries and executables

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



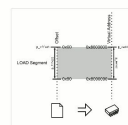
## LOADING PROCESS

### 1 HEADER

THE ELF HEADER IS PARSED  
THE PROGRAM HEADER IS PARSED  
SECTIONS ARE NOT USED

### 2 MAPPING

THE FILE IS MAPPED IN MEMORY  
ACCORDING TO ITS SEGMENTS



### 3 EXECUTION

ENTRY IS CALLED  
SYSCALLS ARE ACCESSED VIA  
- SYSCALL NUMBER IN THE R7 REGISTER  
- CALLING INSTRUCTION SVC

## TRIVIA

THE ELF WAS FIRST SPECIFIED BY U.S.C. AND U.I.<sup>TM</sup>  
FOR UNIX SYSTEM V, IN 1989

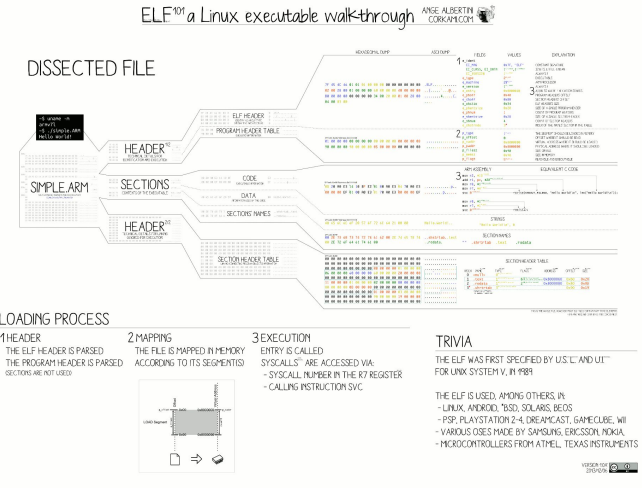
THE ELF IS USED, AMONG OTHERS, IN  
- LINUX, ANDROID, BSD, SOLARIS, BEOS  
- PSP, PLAYSTATION 2-4, DREAMCAST, GAMECUBE, Wii  
- VARIOUS OSes MADE BY SAMSUNG, ERICSSON, NOKIA  
- MICROCONTROLLERS FROM ATMEL, TEXAS INSTRUMENTS



# Tools for inspecting libraries and executables

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

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



# Tools for inspecting libraries and executables

## Practical session

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

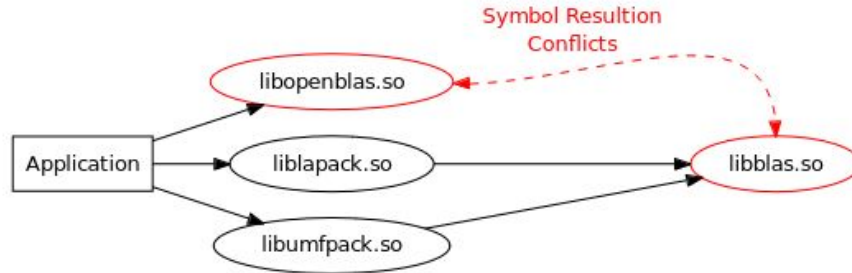


Figure 2: Wrong symbol resolution after relinking the example application.

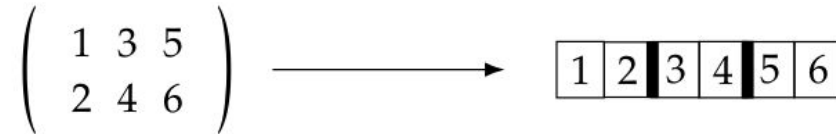


# Data representation for vector and matrices

# Data representation

## Matrices

- Computer memory is linear
- Matrices must be linearized:

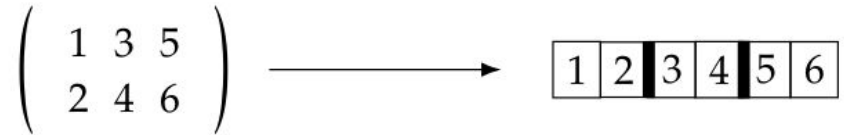




# Data representation

```

structure matrix(T)
  data : T*
  m : integer
  n : integer
end structure
  
```



- How useful is this representation?

# Data representation

## Example: Gaussian elimination algorithm

$$A^{(0)} = \left( \begin{array}{ccc|ccc} 1 & 1 & 1 & 1 & 2 & 2 \\ 1 & 2 & 2 & 1 & 2 & 2 \\ 1 & 2 & 3 & 1 & 2 & 3 \end{array} \right) \sim A^{(1)} = \left( \begin{array}{ccc|ccc} 1 & 1 & 1 & 1 & 2 & 2 \\ 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & 1 & 2 & 0 & 1 & 1 \end{array} \right)$$

$$A^{(0)} \equiv \left( \begin{array}{ccc|ccc|ccc} 1 & 1 & 1 & 1 & 2 & 2 & 1 & 2 & 3 \end{array} \right)$$

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

# Data representation

## Example: Gaussian elimination algorithm

**structure** matrix(T)

data : T\*

ld : integer

m : integer

n : integer

**end structure**

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

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



# The interface of BLAS

# The BLAS interface

---

- 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.

# The BLAS interface

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

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

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

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

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

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

# The BLAS interface

- 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

# The BLAS interface

---

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

## DGEMV:

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

## DGEMM:

- D: double precision
- GE: general matrix
- MM: matrix-matrix multiplication

- The convention does not work always, especially for Level 1 operations
  - DAXPY:  $y = \alpha x + y$



# Data representation & BLAS interface

## Practical session

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



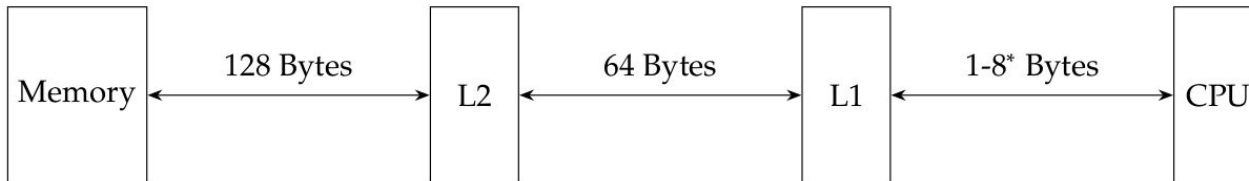
# Effects of caches and cache aware programming

# Effects of caching

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

```

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



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

# Effects of caching

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

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

Vectorizable:



Non-vectorizable:



# Effects of caching

- 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;
}
```

Vectorizable:



Non-vectorizable:



# Effects of caching

- 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;
    a[4*i+3] = 0;
}
  
```

Vectorizable:



Non-vectorizable:



# Effects of caching

---

## 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](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?