



Towards an efficient im- plementation of CADNA in the BLAS : Example of DgemmCADNA routine

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Summary



- 1. Motivations**
- 2. The CADNA Library**
- 3. CADNA Implementation in scientific libraries**
- 4. Conclusion**



1. Motivations

1. Motivations

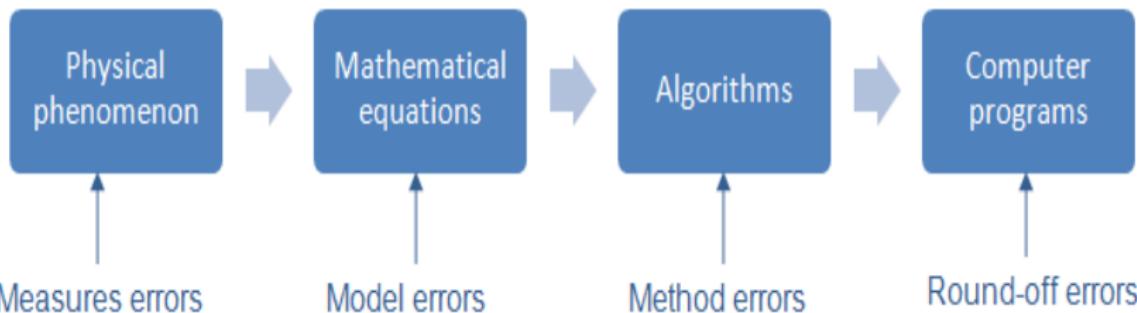
2. The CADNA Library

3. CADNA Implementation in scientific libraries

4. Conclusion

Numerical simulation

Several approximations !



Computed results can be wrong !

- ▶ Round-off errors at each elementary arithmetic operation
- ▶ Detect and control these errors
 - ▶ Numerical validation

Numerical validation : Tools/Methods (1)

► Methods for accurate computations

- ▶ Multiple precision arithmetic :
 - ▶ ex : MPFR, Gnu MP, MPFI.
- ▶ Compensated methods :
 - ▶ compensated summation algorithms, compensated dot product algorithms...

Numerical validation : Tools/Methods (2)

► Methods for rounding error analysis

- ▶ Inverse analysis :
 - ▶ provides error bounds for the computed results.
- ▶ Interval arithmetic : the result of an arithmetic operation between two intervals contains all values that can be obtained by performing this operation on elements from each interval.
- ▶ Probabilistic approach :
 - ▶ uses a random rounding mode (CESTAC Method) ;
 - ▶ estimates the number of exact significant digits of any computed result.

High Performance Computing at EDF R&D

Codes

Code_Aster
Code_Saturne
TELEMAC
Code_...
...

Tools

MPI/OpenMP
BLAS/LAPACK
MUMPS/PASTIX
...
...

Hardware

Ivanoe
Blue Gene
Clamart2
Z600
...



Find an adapted tool for the industrial context (EDF)



The CADNA Library [SJDC07].



2. The CADNA Library

1. Motivations
2. The CADNA Library
3. CADNA Implementation in scientific libraries
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The CESTAC method :

The CESTAC method (Contrôle et Estimation Stochastique des Arrondis de Calculs) was proposed by M. La Porte and J. Vignes in 1974 [VLP74].

It consists in running the same code several times with different round-off error propagations. Then, different results are obtained.

- ▶ the part that is common to all the different results is assumed to be also in common with the mathematical result ;
- ▶ the part that is different in the results is affected by the round-off errors.

Discrete Stochastic Arithmetic

- ▶ N different runs with random rounding mode ($+\infty$ ou $-\infty$ with the probability 0.5) ;
- ▶ N different results R_i :
 - ▶ choosing as the computed result the mean value \bar{R} of R_i ;
 - ▶ estimating C_R the number of exact significant decimal digits of \bar{R} .
- ▶ $N = 3$
 - ▶ $X = (X_1, X_2, X_3)$
 - ▶ $\forall \Omega \in (+, -, \times, /), X \Omega Y = (X_1 \omega Y_1, X_2 \omega Y_2, X_3 \omega Y_3)$
- ▶ If $C_R \leq 0$ or $\forall i, R_i = 0$, a result R is a computed zero (@.0).
- ▶ New order relationships.
- ▶ Discrete Stochastic Arithmetic (DSA).

The CADNA library

- The CADNA library implements Discrete Stochastic Arithmetic. It allows the estimation of round-off error propagation in any scientific program [JCL10].
- More precisely, CADNA enables one to :
 - estimate the numerical quality of any result
 - control branching statements
 - perform a dynamic numerical debugging
 - take into account uncertainty on data.
- CADNA is a library which can be used with Fortran or C++ programs and also with MPI parallel programs. CADNA can be downloaded from <http://www.lip6.fr/cadna>

How to use the Cadna Library

- ▶ CADNA provides two new numerical types, the stochastic types (3 floating point variables x,y,z and a hidden variable acc) :
 - ▶ type (*single_st*) in single precision
 - ▶ type (*double_st*) in double precision.
- ▶ All the operators and mathematical functions are overloaded for these types.
- ▶ To use the library :
 - ① declaration of the CADNA library
 - ② initialization of the CADNA library
 - ③ substitution of the floating point type by stochastic types
 - ④ change of output statements to print stochastic results with their accuracy
 - ⑤ termination of the CADNA library

High Performance Computing at EDF R&D

Codes

Code_Aster

Code_Saturne

TELEMAC

Code_...

...

Tools

MPI/OpenMP

BLAS/LAPACK

MUMPS/PASTIX

...

...

Harware

Ivanoe

Blue Gene

Clamart2

Z600

...

- Is it possible to study the numerical quality of every industrial code with CADNA ?



3. CADNA Implementation in scientific libraries

1. Motivations

2. The CADNA Library

3. CADNA Implementation in scientific libraries

- The communication standards MPI and BLACS
- CADNA implementation in BLAS routines

4. Conclusion

Different extensions for CADNA

- ① MPI extension for CADNA : CADNA MPI
- ② BLACS extension for CADNA : CADNA BLACS
- ③ Efficient implementation of CADNA in BLAS

MPI/BLACS extensions for CADNA

- ▶ Definition of stochastic types to exchange data
- ▶ Definition of reduction operators
- ▶ C/C++ (MPI2) , Fortran 90 (MPI1)

- ▶ (--) The Sendind time of a stochastic float is 4 times more long than a normal float one.
 - ▶ Size of stochastic type = 4 *times* size of normal float
- ▶ (++) It is possible to use CADNA with any code using MPI and BLACS.

BLAS : Basic Linear Algebra Subprograms

► Functionality

- ▶ Level 1 : vectors operations (ex $xAXPY$) ;
- ▶ Level 2 : matrix-vectors operations (ex $xGEMV$) ;
- ▶ Level 3 : matrix-matrix operations (ex $xGEMM$) .

► Implementations

versions	daxpy	dgemv	dgemm
Netlib	1.18482	1.15347	1.35378
Mkl 1 threads	2.02116	2.11232	7.53686
Goto 1 threads	2.86331	2.12331	7.52166
Mkl 8 threads	1.63618	2.79974	58.0523
Goto 8 threads	1.63618	4.60287	56.3343

TABLE: GFLOPS for daxpy, dgemv et dgemm : 4096*4096 matrix (4096 vector).

How to use CADNA with Blas routines ?

- The easiest solution (V1) : Replaced *float* by *float_st* et *double* by *double_st* :

```
void cblas_dgemm(const enum CBLAS_ORDER ←
                  Order,const enum CBLAS_TRANSPOSE ←
                  TransA,const enum CBLAS_TRANSPOSE ←
                  TransB,const int M,const int N,const ←
                  int K,const double_st alpha,const ←
                  double_st *A,const int lda,const ←
                  double_st *B,const int ldb,const ←
                  double_st beta,double_st *C,const int ←
                  ldc);
```

- Linalg : A template version of BLAS ; it can be used with stochastic types

Direct Implementations of DGEMM with CADNA

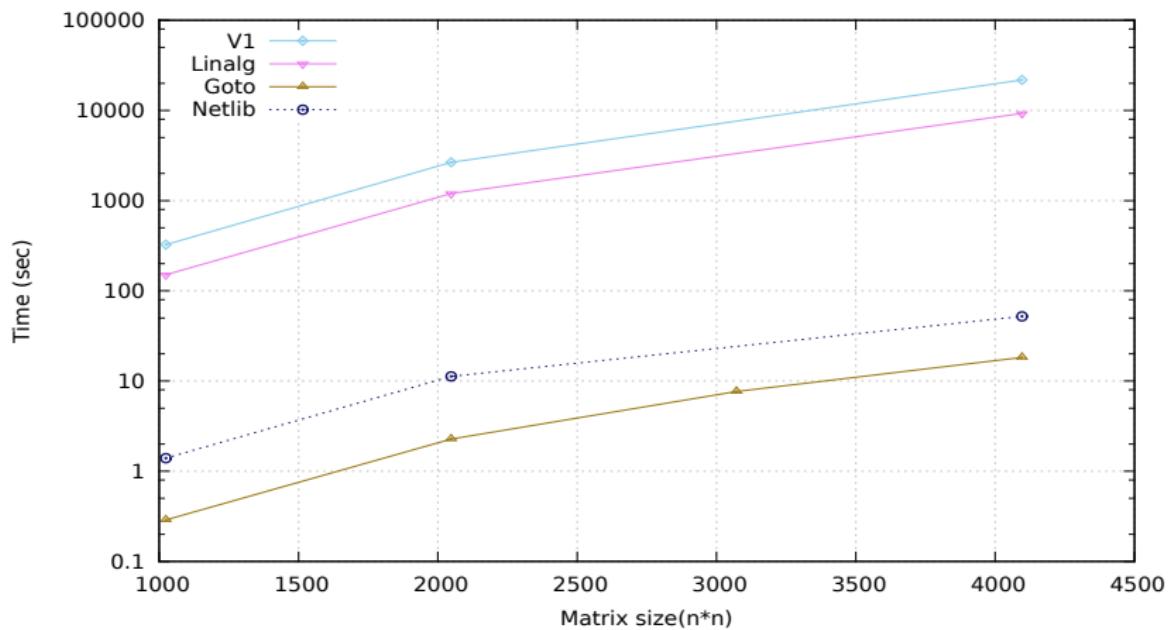


FIGURE: Versions with and without CADNA

Direct Implementations of DGEMM with CADNA(2)

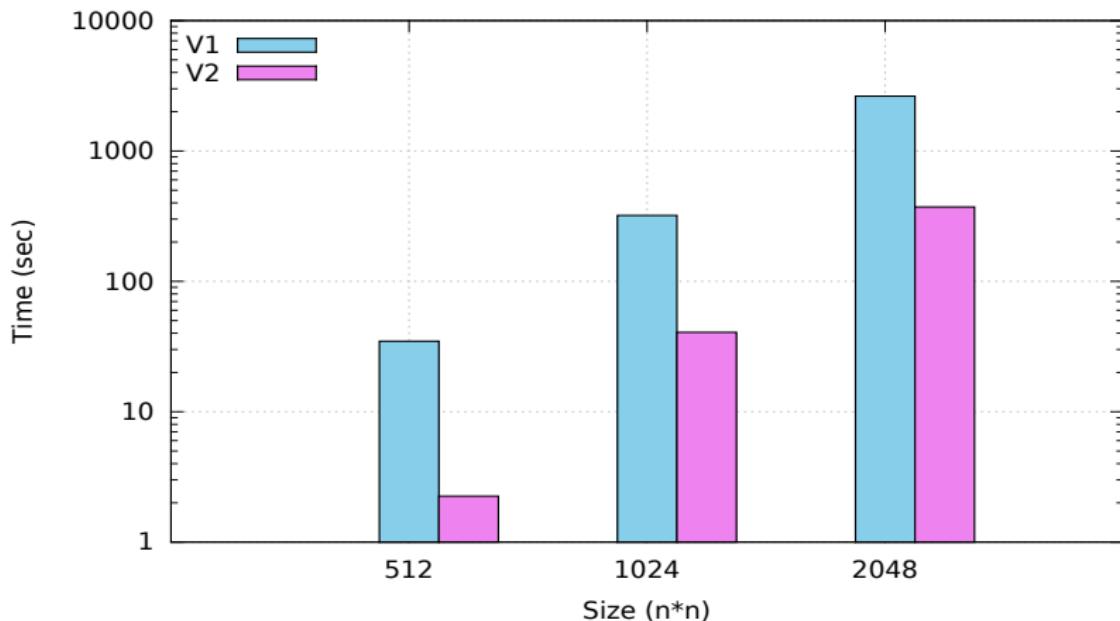


FIGURE: Overhead due to the CESTAC Method

Why these overheads ?

- ▶ An overhead greater than **1000** for a 1024×1024 matrix
 - ▶ DGEMM with 3 inner loops => cache misses
 - ▶ Use of stochastic types and the discrete stochastic arithmetic
 - ▶ Random rounding mode $V1 > 7xV2$

Efficient implementation of DgemmCADNA

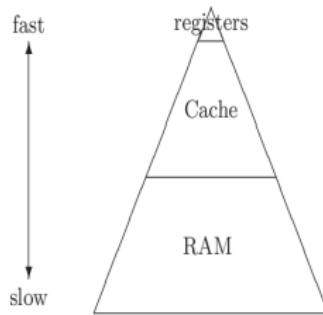
« Implementing matrix multiplication so that near-optimal performance is attained requires a thorough understanding of how the operation must be layered at the macro level in combination with careful engineering of high-performance kernels at the micro level. »

K. Goto, 2008 [GVDG08].

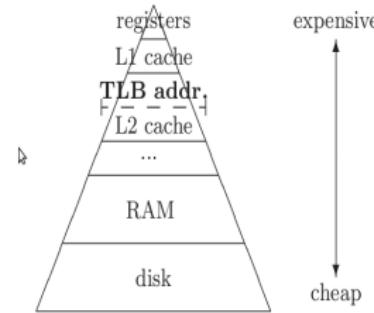
► Solutions to reduce the overhead :

- ① Efficient use of data (memory access)
- ② Minimize the CESTAC Method impact
- ③ Optimize the inner loop

Efficient use of data or data reuse



Simple Model



Refined Model

- ▶ Use tiled algorithms
- ▶ Optimize cache locality
- ▶ Exploit temporal and spacial locality
- ▶ Reduce cache misses
- ▶ Reduce TLB misses

An Iterative tiled algorithm

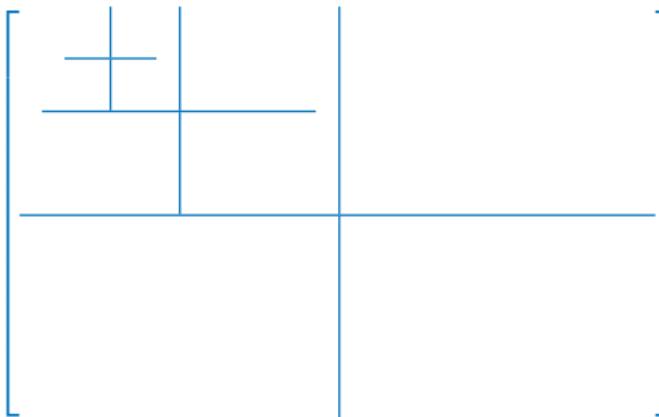
$$\begin{bmatrix} C_{11} & C_{12} & \dots & C_{1N} \\ C_{21} & C_{22} & \dots & C_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ C_{N1} & C_{N2} & \dots & C_{NN} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{N1} & A_{N2} & \dots & A_{NN} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} & \dots & B_{1N} \\ B_{21} & B_{22} & \dots & B_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ B_{N1} & B_{N2} & \dots & B_{NN} \end{bmatrix}$$

every C_{ij} is computed by :

$$C_{ij} = \sum_{k=1}^N A_{ik} B_{kj}$$

A recursive tiled algorithm DGBRn

$$\left[\begin{array}{c|c} C_{00} & C_{01} \\ \hline C_{10} & C_{11} \end{array} \right] = \left[\begin{array}{c|c} A_{00} & A_{01} \\ \hline A_{10} & A_{11} \end{array} \right] \times \left[\begin{array}{c|c} B_{00} & B_{01} \\ \hline B_{10} & B_{11} \end{array} \right] \quad (1)$$



An iterative tiled algorithm based on the hardware (hierarchical memory) DGBIn

► 3 levels of partitioning. One level for every cache level. The matrix (submatrices) is partitioned in submatrices (blocks). At each step, 3 blocks must fit in this level of cache memory.

- ① First level for Cache L3
- ② Second level for Cache L2
- ③ Third level for Cache L1

$$A(n * n) = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{N1} & A_{N2} & \dots & A_{NN} \end{bmatrix}$$

BDL : *Block Data Layout*

► Column Major order

$$M = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 4 & 7 & 2 & 5 & 8 & 3 & 6 & 9 \end{bmatrix}$$

► Row Major order

$$M = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \end{bmatrix}$$

BDL : Block Data Layout (2)

Consider matrix $A(n \times n)$ partitioned in $N \times N$ submatrices A_{ij} :

$$A(n*n) = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{N1} & A_{N2} & \dots & A_{NN} \end{bmatrix} \quad A_{ij}(p*p) = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1p} \\ a_{21} & a_{22} & \dots & a_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{p1} & a_{p2} & \dots & a_{pp} \end{bmatrix}$$

Data within one such block A_{ij} are mapped onto contiguous memory :

$$[a_{11} \ a_{12} \ \dots \ a_{1p} \ a_{21} \ a_{22} \ \dots \ a_{2p} \dots a_{p1} \ a_{p2} \ \dots \ a_{pp}]$$

These blocks are arranged in *row-major order* :

$$[A_{11} \ A_{12} \ \dots \ A_{1N} \ A_{21} \ A_{22} \ \dots \ A_{2N} \dots A_{N1} \ A_{N2} \ \dots \ A_{NN}]$$

Reduce the impact of DSA

Unroll every Cadna arithmetic operation : **NO MORE** operator overloading

```
C[i] = A[i] + B[i] ;
```

```
C[i].x = A[i].x + B[i].x ;
if (random) rnd_switch();
C[i].y = A[i].y + B[i].y ;
if (random) rnd_switch();
C[i].z = A[i].z + B[i].z ;
rnd_switch();
```

Reduce the impact of DSA (2)

less calls to rnd_switch()

```
if(random) rnd_switch()
C[i].x = A[i].x + B[i].x ;
C[i].z = A[i].z + B[i].z ;
C[i+1].z = A[i+1].z + B[i+1].z ;
C[i+2].x = A[i+2].x + B[i+2].x ;
C[i+2].y = A[i+2].y + B[i+2].y ;
C[i+3].x = A[i+3].x + B[i+3].x ;
rnd_switch();
C[i].y = A[i].y + B[i].y ;
C[i+1].x = A[i+1].x + B[i+1].x ;
C[i+1].y = A[i+1].y + B[i+1].y ;
C[i+2].z = A[i+2].z + B[i+2].z ;
C[i+3].y = A[i+3].y + B[i+3].y ;
C[i+3].z = A[i+3].z + B[i+3].z ;
```

Optimize the kernel

Listing 1 – Inner loops

```
for(int i = 0; i < nb_block; i++){
    for(int k = 0; k < nb_block; k++){
        for(int j = 0; j < nb_block; j++){
            Cij = Aik * Bkj /*kernel*/
```

$$C_{00} = A_{00} \times B_{00}$$

$$C_{01} = A_{00} \times B_{01}$$

$$C_{00} = A_{01} \times B_{10}$$

$$C_{01} = A_{01} \times B_{11}$$

$$C_{10} = A_{10} \times B_{00}$$

$$C_{11} = A_{10} \times B_{01}$$

$$C_{10} = A_{11} \times B_{10}$$

$$C_{11} = A_{11} \times B_{11}$$

Results

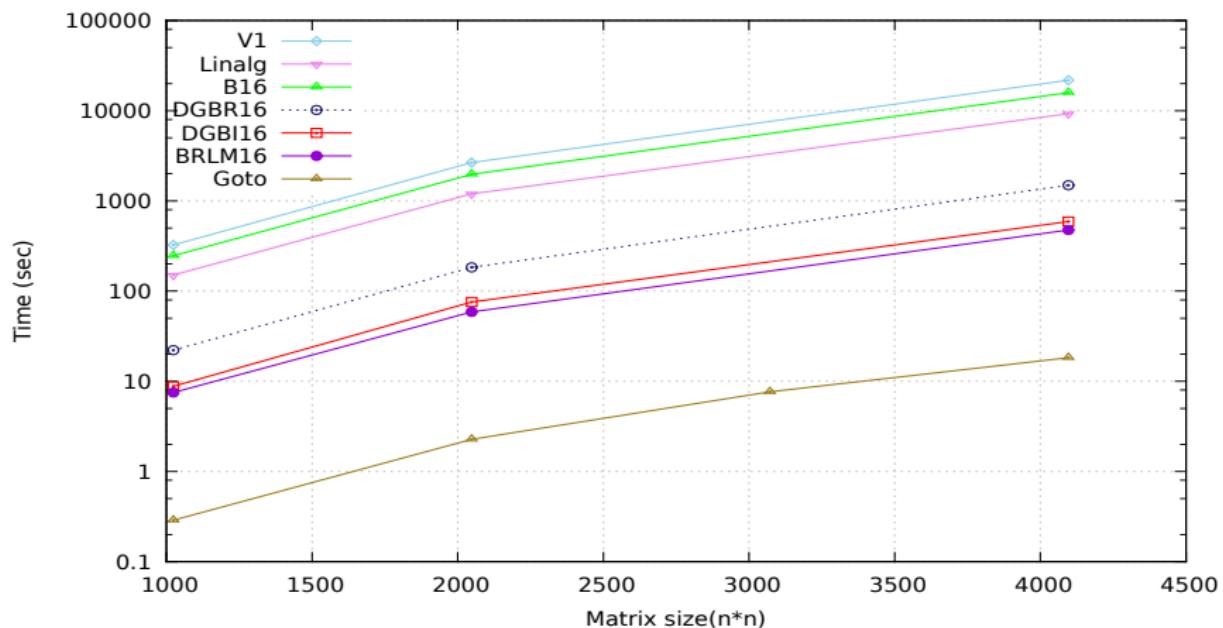


FIGURE: Different versions of DgemmCADNA

Results (2)

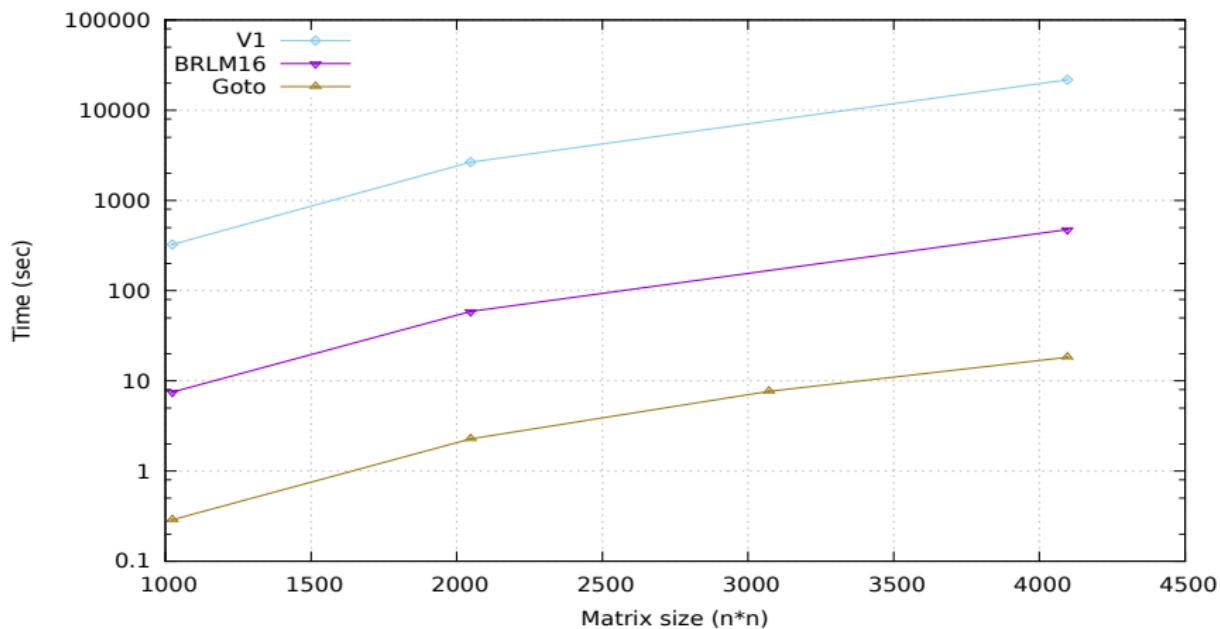


FIGURE: Comparison to GotoBLAS



4. Conclusion

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Conclusion et Future work

► Conclusion

- ▶ CADNA extensions for MPI and BLACS ;
- ▶ DgemmCADNA subroutine :
 - ▶ 45x faster than the first version
 - ▶ Gotoblas 25x faster than DgemmCADNA

► Future work

- ▶ Include the CADNA autovalidation ;
- ▶ Theoretical proof of the CESTAC Method modification
- ▶ Work on the other blas routines (level 1 and level 2)
- ▶ Experimental test phase for the implemented routines in a industrial codes (TELEMAC)

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Thanks !

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