User's Guide for MuPAT on MATLAB

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1 Introduction

The multiple precision arithmetic toolbox MuPAT implemented on Scilab[1] and MATLAB[2] can be used pseudo-quadruple-precision numbers (double-double)[3] and pseudo-octuple-precision numbers (quad-double)[4]. It is prepared for each OS independently (currently, MacOS only). We can easily handle the above high-precision operations interactively using the MATLAB command window, and with same MATLAB codes. Furthermore, MuPAT is applied parallel processing to vector and matrix operations. As the parallel processing features, FMA, AVX2 and OpenMP are used. For more information on parallel processing, please refer to the Parallel processing section.

1.1 System requirements

MuPAT is available in the following environment.

MacOS

MATLAB R2017a or more later

CPU: intel Haswell or more later, amd processor is surveying now...

2 How to start MuPAT

2.1 Install

After download zip file of MuPAT from the web page [https://www.ed.tus.ac.jp/1419521/] or from GitHub [https://github.com/HotakaYagi/MuPAT], unzip and move MuPAT to a directory that is easy to work, such as under the MATLAB directory.

2.2 Start-up

- 1. Launch MATLAB
- 2. Move current directory to MuPAT
- 3. Enter the following command to the MATLAB command window

>> startMuPAT(n1, n2, n3)

After completing the above steps, enter the following command to the MATLAB command window, and if the following result is output, MATLAB is ready to use MuPAT.

```
>> a = DD(1)
a =
```

hi: 1 lo: 0

MuPAT can select the number of threads, valid or invalid of AVX2, valid or invalid of FMA as input arguments. (No arguments means the number of threads = 1, FMA, and AVX2 are set off)

Table 2.1: The input arguments for startMuPAT

	the number of threads (n1)	AVX2 (n2)	FMA (n3)
Speed up ON	$\{n \mid n \in \mathbb{N}, n > 1\}$	1	1
Speed up OFF	1 (serial) or No arguments	0 or No arguments	0 or No arguments

Ex.

startMuPAT: Computing under 1 threads, invalid of AVX2, and invalid of FMA (serial computing) startMuPAT(2, 1, 1): Computing under 2 threads, valid of AVX2, and vaild of FMA.

2.3 Remarks

Be aware of the required memory size for the input you are computing. MATLAB MEX function may kill MATLAB if the requested memory size is large. Furthermore, the computation order differs from the serial computing and parallel computing. It means they have different rounding errors. As a result, the computation results may not match completely as serial computing.

3 Usage

In MuPAT, we can use quadruple and octuple precision arithmetics in the same way as double precision arithmetic, because MATLAB can define new data type and apply overloading for operators and some MATLAB functions. The special command in MuPAT is following.

- The declaration or conversion of variables of type DD and QD (section 3.1 and section 3.2).
- The ways of input and output (section 3.3).
- Some of DD and QD function (Section 3.4).

3.1 Variable definition

We define new data types named DD and QD to contain double-double and quad-double numbers. These types require two or four double precision numbers. The type DD has two double numbers of hi (upper part) and 1o (lower part) for quadruple precision, and type QD has four double numbers of hh (the highest part), hl, 1h and 1l (the lowest part) for octuple precision. The upper part of DD variable A can be referenced with A.hi, and the lower is with A.lo. QD variable B can be referenced with B.hh, B.hl, B.lh, and B.ll from the highest part to the lowest part respectively.

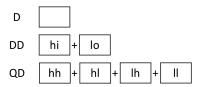


Figure 3.1: Images of type DD and type QD

To define type DD, you need to use the following function. The DD function need up to two arguments. The type of arguments does not matter.

>> DD(a, b)

- DD(), if there is no argument, the value ois assigned to the upper part and the lower part.
- DD(a), if only one argument is input, a is assigned to the upper part and the lower part is assigned the value 0
- When two arguments are input, the input values are assigned to upper part and lower part respectively.

To define type QD, you need to use the following function. The QD function need up to four arguments. The type of arguments does not matter.

>> QD(a, b, c, d)

- QD(), if there is no argument, the value 0 is assigned to from the highest part to the lowest part.
- QD(a), if only one argument is input, a is assigned to the highest part, and the value 0 is assigned to the others.
- QD(a, b), if two arguments are input, a and b are assigned to the upper two places, and the value 0 is assigned to the lower two places.
- QD(a, b, c), if there are three arguments, a, b and c are assigned to the upper three places, and value 0 is assigned to the lowest part.

• When four arguments are input, the input values ina, b, c and d are assigned to from the highest part to the lowest part respectively.

The components of DD and QD are substituted in descending (renormalized) order.

```
\mathbf{E}\mathbf{x}.
>>a = DD(1, 0)
hi:
     1
lo: 0
>> b = QD(3.14, 0, 0, 0)
hh:
    3.14
hl:
     0
lh:
     0
11:
>> A = DD([1, 2, 3; 4, 5, 6])
A =
     [2x3 double]
hi:
     [2x3 double]
>> A.hi
ans =
1 2 3
4 5 6
>> A.lo
ans =
0 0 0
0 0 0
```

3.2 Type conversion

We can convert the variables of data types (double, DD, or QD) into different data types (double, DD, or QD) with the following functions. Of course it is the same way even if a is a vector or matrix.

Table 3.1: The way to type conversion

from\to	double	DD	QD	
double		DD(a)	QD(a)	
DD	double(b)		QD(b)	
QD	double(c)	DD(c)		

$\mathbf{E}\mathbf{x}$.

To convert variable b of type DD into type double:

>> double(b)

To convert variable a type double into type QD:

>> QD(a)

3.3 Input/Output

MuPAT has the following input and output functions. These functions can only use scalar variables.

- ddinput(s)
 - Converts a numeric string s into numeric value of type DD.
- qdinput(s)

Converts a numeric string s into numeric value of type QD.

- ddprint(a)
 Display the variable of type DD as a character string in 31 decimal places.
- qdprint(a)
 Display the variable of type QD as a character string in 63 decimal places.

3.4 Other DD and QD functions

A random DD and QD numbers are generated by the MATLAB function. The upper part and the lower part in the DD are assigned random numbers and are normalized to one value. In the same way, from the highest to the lowest part of QD numbers are assigned random numbers and normalized to one value.

ddrand(m, n)
Returns a pseudorandom DD matrix that size is m by n.
qdrand(m, n)
Returns a pseudorandom QD matrix that size is m by n.
ddeye(m, n)
Returns an identity matrix of type DD that size is m by n.
qdeye(m, n)
Returns an identity matrix of type QD that size is m by n.

3.5 Arithmetic operators

3.5.1 Four arithmetic operators

We can use the same operators (+, -, *, /) in type DD and QD, even if variables are vectors or matrices. (Division operator is only supported in scalars.) We can also compute the sum of variables of type DD and double using the operator +. In other words, mixed precision arithmetic is also available. When we do mixed precision arithmetic of type DD and double, they will be converted to type DD in the computation. The same rule applies to type QD. It is like an implicit type conversion between int and double on usual programming language.

```
Ex. 1
>> d = 1 + 1.0E-20
d =
1
>> e = DD(1, 1.0E-20);
hi:
   1
lo: 1.0000e-20
>> e - d
ans =
   1.0000e-20
hi:
lo:
   0
Ex. 2
>> format long
>> disp(1/3);
0.3333333333333333
>> ddprint(DD(1)/3);
' 3.3333333333333333333333333333E-1'
>> qdprint(QD(1)/3);
```

X If the command window is small, some of the displayed output values may be omitted.

3.5.2 Dot operators

We can use the same dot operations (.*, ./) in type DD and QD as double.

```
Ex. 3
>> A = DD([1.1, 1.2, 1.3]);
>> B = DD([1.1, 1.2, 1.3]);
>> C = A .* B
hi: [1.2100 1.4400 1.6900]
lo: [8.8818e-18 -5.3291e-17 -5.3291e-17]

Ex. 4
>> A = DD([1.1, 1.2, 1.3]);
>> B = DD([1.1, 1.2, 1.3]);
>> C = A ./ B
ans =
hi: [1 1 1]
lo: [0 0 0]
```

3.6 Relational operators

We can use the same operator $(=, \sim =, <, >, \le, \ge)$ for type DD and QD as double. These operators return logical value 0 or 1.

```
Ex. 5
>> a = 1/7;
>> b = 1/DD(7);
>> c = 1/QD(7);
>> a == b
ans =
logical
>> a < b
ans =
logical
>> a \sim= c
ans =
logical
>> a == b.hi
ans =
logical
```

3.7 Logical operators

We can use the same operator $(\&,|,\sim)$ for type DD and QD as double. These operators return logical value 0 or 1.

```
Ex. 6
>> A = DD([1,0,3; 4,5,0]);
>> B = DD([0,0,3;4,0,6]);
>> A & B
ans =
```

```
2x3 logical array
0 0 1
1 0 0
>> A | B
ans =
2x3 logical array
1 0 1
1 1 1
>> ~A
ans =
2x3 logical array
0 1 0
0 0 1
```

3.8 Element insertion and extraction

• A(i, j) = b

Insert the variable b to the (i, j) element of array A. The assignment is not reflected if the precision on the right side is higher than the left side.

```
Ex.
>> A = DD([1,2,3; 4,5,6]);
>> A(2, 3)
ans =
hi: 6
lo: 0
>> A(2, 2) = 10;
>> A.hi
ans =
1 2 3
4 10 6
>> A.lo
ans =
0 0 0
0 0
>> A(2, 2) = QD(1); This is error because A is type DD although right side is type QD.
```

3.9 Other MATLAB functions

3.9.1 Supported functions

• A'

Return the transposed matrix of A.

• [A, B]

Concatenates B horizontally to the end of A when A and B have compatible sizes.

• [A; B]

Concatenates B vertically to the end of A when A and B have compatible sizes.

• [m, n] = size(a)

Return the row length m and the column length n of a.

• abs(a)

Return the absolute value of matrix a.

• sqrt(a)

Return the square root of matrix a.

• dot(x, y)

Return the inner product of vectors ${\tt x}$ and ${\tt y}$.

- tmv(A, b)
 - Return the fast transposed matrix vector multiplication of A' and b.
- a^n

Returns the value of the scalar a raised to the nth power.

• norm(a,n)

Return the nth norm of a. The second argument can be 1, 2, inf or fro.

3.9.2 Unsupported functions

MuPAT does not support functions other than those described in Section 3.9.1. If you call an unsupported function:

```
Ex.
>> a=ddrand(5);
>> qr(a)
Undefined function 'qr' for input arguments of type 'DD'.
```

3.10 The difference from MuPAT on Scilab

Regarding the operation, when declaring type DD and type QD in the Scilab, they were described as dd and qd in small letters, but in MATLAB, it is necessary to call functions in capital letter. In Scilab, when converting types, it was necessary to call the d2dd() function when you want to convert from double to DD, and the d2qd() function when you want to convert from double to type QD. However, in MATLAB, you can convert just do assign a variable to the type you want to convert (see section 3.2). In addition, it was necessary to call the getHi() function in Scilab version when extracting double precision elements such as upper part of type DD and type QD. However, in MATLAB version, just do write a.hi or b.hh in code can extract double elements. The sparse matrix form and the applications such as numerical methods and matrix decomposition (see section 3.9.1 and 3.9.2) are not supported in MATLAB.

4 Parallel processing

To be valid or invalid of FMA, AVX2, and OpenMP, enter the command as follows to the MATLAB command window.

```
>> startMuPAT(n1, n2, n3)
```

Please specify the arguments n1 (the number of threads), n2 (valid or invalid of AVX2) and n3 (valid or invalid of FMA). If you do not specify the input arguments, that will be invalid.

The argument n1 can be a nonnegative input, and n2 and n3 can be 0 (invalid) or 1 (valid).

We can also set valid or invalid as following by entering to the MATLAB command window.

- >> fmaon
- >> fmaoff
- >> avxon
- >> avxoff
- >> omp(n1)

When you check which parallelization is enabled or disabled, you can enter the following command to the MATLAB command window.

>> statusMuPAT

Ex.

- >> startMuPAT
- >> statusMuPAT

Cureently MuPAT is running at the number of threads: 1thread (serial), AVX is off, and FMA is off

>> avxon

```
>> omp(4) Cureently MuPAT is running at the number of threads: 4, AVX is on, and FMA is off
```

4.1 FMA

A floating point multiply-add operation $(x = a \times b + c)$ is performed in one step via a single rounding FMA (fused multiply-add) instruction[5, 6]. As a result, the number of multiply-add operations are reduced up to two times. Multiply-subtraction is also supported. If we set valid of FMA, twoprod_fma[4] routine is called for DD and QD algorithm. Therefore, the internal algorithm in multiply operation differs depending on whether you use FMA or not.

4.2 AVX2

Since four double precision data are processed simultaneously with the AVX2 (Advanced Vector Extensions)[5, 6] instruction, we can expect the computation time would become four times faster.

Ex. >> startMuPAT(1, 0, 0) >> a = qdrand(500); >> b = qdrand(500); >> tic; a * b; toc; Elapsed time is 8.374103 seconds. >> avxon >> tic; a * b; toc; Elapsed time is 2.227658 seconds.

4.3 OpenMP

Since OpenMP[7] enables thread level parallelism within a shared memory, we can expect the computation time to decrease by a multiple of the number of cores. The number of threads indicates how many parallel to compute. The maximum number of threads varies depending on the CPU. The number of threads is not equal to the speedup rate. We can select from 1 to the maximum number of threads. It is usually better to set the number of threads = the number of cores.

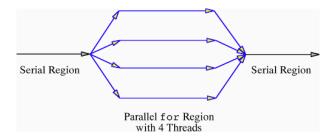


Figure 4.1: Image of multi threads using OpenMP

```
Ex.
>> startMuPAT(1, 0, 0)
>> a = ddrand(500);
>> b = ddrand(500);
>> tic; a * b; toc;
Elapsed time is 0.814772 seconds.
>> omp(2)
set the number of threads = 2
```

```
>> tic; a * b; toc;
Elapsed time is 0.449846 seconds.
```

4.4 Acceleration-supported operations

Using these parallelization, if the data is large enough and the operation count is large, it can be computed 16 to 17 times faster on modern 4-core machine. In the table, 'D-DD' means that a operation $x \circ y$ is performed with double precision variable x and DD variable y, and 'DD-D' mean that a operation $y \circ x$ is performed with DD variable y and double precision variable x. The way to call functions in the table means that the left side of the arrow is input type for the operation $y \circ x$ and the right side of the arrow is the required command for the operation $y \circ x$. The following table shows the 48 of operations are parallelized.

Table 4.1: Parallelization supported functions and how to call them

	D-DD	D-QD	DD-D	DD-DD	DD-QD	QD-D	QD-DD	QD-QD
Vector (matrix) addition	x, y : vector or matrix $\rightarrow x + y$ or $x - y$							
Scalar multiplication	a: scaler, x: vector or matrix \rightarrow a * x or x * a							
Inner product	$x, y: vector \rightarrow dot(x, y) or x' * y$							
Matrix-vector								
multiplication	A: matrix, b: vector \rightarrow A * b							
Transposed matrix-								
vector multiplication	A: matrix, b: vector \rightarrow tmv(A, b)							
Matrix-matrix								
multiplication	A, B: $matrix \rightarrow A * B$							

^{*} Sparse matrix form is not supported.

References

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- [3] D. H. Bailey, High-Precision Floating-point arithmetic in scientific computation, Computing in Science and Engineering, Vol.7(3), pp.54-61
- [4] Y. Hida, X. S. Li and D. H. Baily, Quad-double arithmetic: algorithms, implementation and application, Technical Report LBNL-46996, Lawrence Berkeley National Laboratory, Berkeley (2000).
- [5] Intel: Intrinsics Guide, available from https://software.intel.com/sites/landingpage/IntrinsicsGuide/
- [6] Intel: 64 and IA-32 Architectures Optimization Reference Manual, https://www.intel.com/content/dam/www/public/us/en/documents/manuals/64-ia-32-architectures-optimization-manual
- [7] OpenMP: http://www.openmp.org/
- [8] Matlab Documentation: https://jp.mathworks.com/help/matlab/ref/mex.html

Appendix

A: Contents of toolbox

The folder structure of MuPAT is described below.

□ MuPAT
🗇 README.md
ldot useguideJap.pdf
$ldsymbol{oxtlesh}$ UseguideEng.pdf
🗀 @DD
f DD.m
🗀 @QD
QD.m
🗀 @Calc
☐ Calc.m
□ src
□ ddeye.m
□ qdeye.m
avxoff.m
△ avxon.m
fmaoff.m
fmaon.m
omp.m
statusMuPAT.m
mupat.c
mupat.h
(set of C program)
[](set of MEX files)