Performance evaluation of an efficient double-double BLAS1 function with error-free transformation and its application to explicit extrapolation methods

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# Outline

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# Summary

- For initial value problems of ordinary differential equations (ODEs), we want to obtain more precise double precision numerical solutions more quickly than when using double-double (DD) precision arithmetic.
- We have implemented lighter and accurate BLAS1 functions with EFT and used them to explicit extrapolation methods.

#### ₩

The presented routines can be effective for a large system of linear ODE and for small nonlinear ODE, especially when a harmonic sequence is used.

Initial value problem of ordinary differential equation

Initial value problem of Ordinary Differential Equation (ODE for short) to be solved:

$$\frac{d\mathbf{y}}{dt} = \mathbf{f}(t, \mathbf{y}) \in \mathbb{R}^{n}$$

$$\mathbf{y}(0) = \mathbf{y}_{0}$$
Integration interval :  $[0, t_{end}]$ 

$$\downarrow$$

$$\mathbf{y}$$

$$\mathbf{y}$$

$$\mathbf{y}$$

$$\mathbf{y}$$

$$\mathbf{y}$$

We compute  $\mathbf{y}_{next} \approx \mathbf{y}(t_{next})$  at each  $t_{next} \in [0, t_{end}]$  from  $\mathbf{y}_{old} \approx \mathbf{y}(t_{old})$ .

# Extrapolation for ODE: Bulirsch-Stoer Algorithm

Give a support sequence  $\{w_i\}$ , max. number of stages L, relative tolerance  $\varepsilon_R$  and absolute tolerance  $\varepsilon_A$ . Support sequences:

Romberg:  $2, 4, 8, \dots, 2^i, \dots \Rightarrow$  Stable but Slow

Harmonic:  $2, 4, 6, 8, ..., 2(i+1), ... \Rightarrow$  Unstable but Fast

Process to calculate initial sequence:  $T_{i1}$  (i = 1, 2, ..., L):

1. 
$$h := (t_{\text{next}} - t_{\text{old}})/w_i \longrightarrow t_k := t_{\text{old}} + kh \in [t_{\text{old}}, t_{\text{next}}]$$

- 2.  $t_0 := t_{\text{old}}, y_0 \approx y(t_0)$
- 3. Explicit Euler Method

$$\mathbf{y}_1 := \mathbf{y}_0 + h\mathbf{f}(t_0, \mathbf{y}_0)$$

4. Explicit midpoint method to get  $\mathbf{y}_2$ ,  $\mathbf{y}_3$ , ...  $\mathbf{y}_{w_i}$ 

$$\mathbf{y}_{k+1} := \mathbf{y}_{k-1} + 2h\mathbf{f}(t_k, \mathbf{y}_k) \ (k = 1, 2, ..., w_i - 1)$$

5. Set the initial sequence for extrapolation:  $\mathbf{S}(h/w_i) := \mathbf{y}_{w_i}$ 

Extrapolation for ODE: Bulliursh-Stoer Algorithm (cont.)

1. 
$$\mathbf{T}_{11} := \mathbf{S}(h/w_1)$$
  
2.  $i = 2, ..., L$   
 $\mathbf{T}_{i1} := \mathbf{S}(h/w_i)$   
For  $j = 2, ..., i$ 

Extrapolation to get better approximation:

$$\begin{split} \mathbf{R}_{ij} &:= \left( \left( \frac{w_i}{w_{i-j+1}} \right)^2 - 1 \right)^{-1} \left( \mathbf{T}_{i,j-1} - \mathbf{T}_{i-1,j-1} \right) \\ \mathbf{T}_{ij} &:= \mathbf{T}_{i,j-1} + \mathbf{R}_{ij} \end{split}$$

Check convergence status if :

$$\|\mathbf{R}_{ij}\| \le \varepsilon_R \|\mathbf{T}_{i,j-1}\| + \varepsilon_A \\ \longrightarrow \mathbf{y}_{\text{next}} := \mathbf{T}_{ij}$$
(2)

3.  $\mathbf{y}_{next} := \mathbf{T}_{LL}$  if not converge

# Application of EFT: FMA with error

cf. S.Boldo & J-M. Muller

$$(s, e_1, e_2) := \mathsf{FMAerror}(a, x, y)$$
  

$$s := \mathsf{FMA}(a, x, y) = ax + y$$
  

$$(u_1, u_2) := \mathsf{TwoProd}(a, x)$$
  

$$(\alpha_1, \alpha_2) := \mathsf{TwoSum}(y, u_2)$$
  

$$(\beta_1, \beta_2) := \mathsf{TwoSum}(u_1, \alpha_1)$$
  

$$\gamma := \beta_1 \ominus s \oplus \beta_2$$
  

$$(e_1, e_2) := \mathsf{QuickTwoSum}(\gamma, \alpha_2)$$
  
return  $(s, e_1, e_2)$ 

$$s + e_1 + e_2 = ax + y$$
  
where  $s = a \otimes x \oplus y$   
 $|e_1 + e_2| = \frac{1}{2}\mathbf{u}|s|$  (u is unit of round-off error)  
 $|e_2| = \frac{1}{2}\mathbf{u}|e_1|$ 

Application of EFT2: FMA with error approximated

cf. S.Boldo & J-M. Muller

$$(s, e) := \mathsf{FMAerrorApprox}(a, x, y)$$
  

$$s := \mathsf{FMA}(a, x, y)$$
  

$$(u_1, u_2) := \mathsf{TwoProd}(a, x)$$
  

$$(\alpha_1, \alpha_2) := \mathsf{TwoSum}(y, u_1)$$
  

$$\gamma := \alpha_1 \ominus s$$
  

$$e := (u_2 \oplus \alpha_2) \oplus \gamma$$
  
return  $(s, e)$ 

When IEEE754 double precision arithmetic is used in FMAerrorApprox, the error bound is provided as  $|(s+e) - (ax+b)| \le 7 \cdot 2^{-105} |s|.$ 

### Application of EFT: BLAS1 with error

$$\begin{aligned} \mathbf{y} &:= \mathsf{AXPY}(\alpha, \mathbf{x}, \mathbf{y}) \\ \mathbf{y} &:= \alpha \otimes \mathbf{x} \oplus \mathbf{y} \\ \textbf{return } \mathbf{y} \end{aligned}$$

∜

# $\begin{aligned} &(\mathbf{y}, \mathbf{e}_{\mathbf{y}}) := \mathsf{AXPYerror}(\alpha, e_{\alpha}, \mathbf{x}, \mathbf{e}_{\mathbf{x}}, \mathbf{y}, \mathbf{e}_{\mathbf{y}}) \\ &(\mathbf{y}, \mathbf{e}_{1}, \mathbf{e}_{2}) := \mathsf{FMAerror}(\alpha, \mathbf{x}, \mathbf{y}) \\ &\mathbf{e}_{\mathbf{y}} := \mathbf{e}_{1} \oplus \mathbf{e}_{2} \oplus \alpha \otimes \mathbf{e}_{\mathbf{x}} \oplus e_{\alpha} \otimes \mathbf{x} \oplus \mathbf{e}_{\mathbf{y}} \\ &\mathbf{return} \ (\mathbf{y}, \mathbf{e}_{\mathbf{y}}) \end{aligned}$

or

$$\begin{aligned} & (\mathbf{y}, \mathbf{e}_{\mathbf{y}}) := \mathsf{AXPYerrorA}(\alpha, e_{\alpha}, \mathbf{x}, \mathbf{e}_{\mathbf{x}}, \mathbf{y}, \mathbf{e}_{\mathbf{y}}) \\ & (\mathbf{y}, \mathbf{e}) := \mathsf{FMAerrorApprox}(\alpha, \mathbf{x}, \mathbf{y}) \\ & \mathbf{e}_{\mathbf{y}} := \mathbf{e} \oplus \alpha \otimes \mathbf{e}_{\mathbf{x}} \oplus e_{\alpha} \otimes \mathbf{x} \oplus \mathbf{e}_{\mathbf{y}} \\ & \mathbf{return} \ (\mathbf{y}, \mathbf{e}_{\mathbf{y}}) \end{aligned}$$

# Application of EFT: BLAS1 with error

$$\mathbf{x} := \mathsf{SCAL}(\alpha, \mathbf{x})$$
$$\mathbf{x} := \alpha \otimes \mathbf{x}$$
return x

#### $\Downarrow$

$$\begin{aligned} & (\mathbf{x}, \mathbf{e}_{\mathbf{x}}) := \mathsf{SCALerror}(\alpha, e_{\alpha}, \mathbf{x}, \mathbf{e}_{\mathbf{x}}) \\ & (\mathbf{w}_1, \mathbf{w}_2) := \mathsf{TwoProd}(\alpha, \mathbf{x}) \\ & \mathbf{w}_2 := \alpha \otimes \mathbf{e}_{\mathbf{x}} \oplus e_{\alpha} \otimes (\mathbf{x} \oplus \mathbf{e}_{\mathbf{x}}) \oplus \mathbf{w}_2 \\ & (\mathbf{x}, \mathbf{e}_{\mathbf{x}}) := \mathsf{QuickTwoSum}(\mathbf{w}_1, \mathbf{w}_2) \\ & \mathsf{return} \ (\mathbf{x}, \mathbf{e}_{\mathbf{x}}) \end{aligned}$$

#### Extrapolation with EFT

Approximation  $\Longrightarrow$  (Approximation, its error)  $\mathbf{f}(t_k, \mathbf{y}_k) := \mathbf{f}_k \implies \mathbf{f}(t_k + e_{t_k}, \mathbf{y}_k + \mathbf{e}_{\mathbf{y}_k}) = \mathbf{f}_k + \mathbf{e}_{\mathbf{f}_k}$ Explicit Euler Method

$$\begin{aligned} \mathbf{y}_1 &:= \mathbf{y}_0 + h\mathbf{f}_0 \\ & \downarrow \\ (\mathbf{y}_1, \mathbf{e}_{\mathbf{y}_1}) &:= (\mathbf{y}_0, \mathbf{e}_{\mathbf{y}_0}) \\ (\mathbf{y}_1, \mathbf{e}_{\mathbf{y}_1}) &:= \mathsf{AXPYerror}(h, e_h, \mathbf{f}_0, \mathbf{e}_{\mathbf{f}_0}, \mathbf{y}_1, \mathbf{e}_{\mathbf{y}_1}) \\ & \mathsf{or} &:= \mathsf{AXPYerrorA}(h, e_h, \mathbf{f}_0, \mathbf{e}_{\mathbf{f}_0}, \mathbf{y}_1, \mathbf{e}_{\mathbf{y}_1}) \end{aligned}$$

Explicit midpoint method

$$\mathbf{y}_{k+1} := \mathbf{y}_{k-1} + 2h\mathbf{f}_k \ (k = 1, 2, ..., w_i - 1)$$

$$\downarrow$$

$$(\mathbf{y}_{k+1}, \mathbf{e}_{\mathbf{y}_{k+1}}) := (\mathbf{y}_{k-1}, \mathbf{e}_{\mathbf{y}_{k-1}})$$

$$(\mathbf{y}_{k+1}, \mathbf{e}_{\mathbf{y}_{k+1}}) := \mathsf{AXPYerror}(2 \otimes h, 2 \otimes e_h, \mathbf{f}_k, \mathbf{e}_{\mathbf{f}_k}, \mathbf{y}_{k+1}, \mathbf{e}_{\mathbf{y}_{k+1}})$$

$$\mathsf{or} := \mathsf{AXPYerror}(2 \otimes h, 2 \otimes e_h, \mathbf{f}_k, \mathbf{e}_{\mathbf{f}_k}, \mathbf{y}_{k+1}, \mathbf{e}_{\mathbf{y}_{k+1}})$$

$$(k = 1, 2, ..., w_i - 1)$$

# Extrapolation with EFT (cont.)

Extrapolation Process Preliminary (DD):  $(c_{ij}, e_{c_{ij}}) := 1/((w_i/w_{i-j+1})^2 - 1)$ 

$$\begin{split} (\mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) &:= (\mathbf{T}_{i,j-1}, \mathbf{e}_{\mathbf{T}_{i,j-1}}) \\ (\mathbf{T}_{ij}, \mathbf{e}_{\mathbf{T}_{ij}}) &:= (\mathbf{T}_{i,j-1}, \mathbf{e}_{\mathbf{T}_{i,j-1}}) \\ (\mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) &:= \mathsf{AXPYerror}(-1, 0, \mathbf{T}_{i-1,j-1}, \mathbf{e}_{\mathbf{T}_{i-1,j-1}}, \mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) \\ & \mathsf{or} &:= \mathsf{AXPYerror}\mathsf{A}(-1, 0, \mathbf{T}_{i-1,j-1}, \mathbf{e}_{\mathbf{T}_{i-1,j-1}}, \mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) \\ (\mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) &:= \mathsf{SCALerror}(c_{ij}, e_{c_{ij}}, \mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}) \\ (\mathbf{T}_{ij}, \mathbf{e}_{\mathbf{T}_{ij}}) &:= \mathsf{AXPYerror}(1, 0, \mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}, \mathbf{T}_{ij}, \mathbf{e}_{\mathbf{T}_{ij}}) \\ & \mathsf{or} &:= \mathsf{AXPYerror}\mathsf{A}(1, 0, \mathbf{R}_{ij}, \mathbf{e}_{\mathbf{R}_{ij}}, \mathbf{T}_{ij}, \mathbf{e}_{\mathbf{T}_{ij}}) \end{split}$$

#### Møller method

The Møller method is proposed to reduce the accumulation of round-off errors incurred during the approximation of IVPs of ODEs and is a type of compensated summation. For the original summation  $S_i := S_{i-1} + z_{i-1}$ , we compute it as follows:

$$s_i := z_{i-1} \ominus R_{i-1} (R_0 = 0)$$
  

$$S_i := S_{i-1} \oplus s_i$$
  

$$r_i := S_i \ominus S_{i-1}$$
  

$$R_i := r_i \ominus s_i.$$

∜

$$s_{i} := z_{i-1} \oplus R'_{i-1} \ (R'_{0} = 0)$$
  
(S<sub>i</sub>, R'<sub>i</sub>) := QuickTwoSum(S<sub>i-1</sub>, s<sub>i</sub>). (3)

### Computing environment

# **Ryzen** AMD Ryzen 1700 (2.7 GHz), Ubuntu 16.04.5, GCC 5.4.0, QD 2.3.18[7], LAPACK 3.8.0.

Corei7 Intel Core i7-9700K (3.6GHz), Ubuntu 18.04.2, GCC 7.3.0, QD 2.3.20, LAPACK 3.8.0.

# Targetted algorithms

Our targets of precision are IEEE754 double precision (Double) and DD provided by the QD library. The targeted algorithms are as follows:

DEFT	Double precision and AXPYerror
DEFTA	Double precision, $\mathbf{f} + \mathbf{e_f}$ , and AXPYerrorA
DMøller	Double precision Møller method.

DEFTA means the usage of the FMA errorA in the entire extrapolation process. For DEFT, DEFTA and DD computations, we used DD precision  ${\bf f}.$ 

$$\|\mathbf{R}_{ij}\| \le \varepsilon_R \|\mathbf{T}_{i,j-1}\| + \varepsilon_A \tag{4}$$

we used  $\varepsilon_R = \varepsilon_A = 0$  unless otherwise specified.

All EFT basic functions were coded as C macros.

# Numerical experiments

1.

$$\frac{d}{dt} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} -y_1 \\ \vdots \\ -ny_n \end{bmatrix} \Longrightarrow \mathbf{y}(t) = \begin{bmatrix} \exp(-t) \\ \vdots \\ \exp(-nt) \end{bmatrix}$$
$$\mathbf{y}(0) = \begin{bmatrix} 1 \ 1 \ \cdots \ 1 \end{bmatrix}^T, t \in [0, 1/4], n = 2048.$$

2.

$$\frac{d}{dt} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} y_2 \\ -\alpha y_1^2 \sin t + 2\alpha y_1 y_2 \cos t \end{bmatrix}$$
$$\mathbf{y}(0) = \begin{bmatrix} 1 \ \alpha \end{bmatrix}^T, \ t \in [0, 37]$$

where  $\alpha=0.999999999$ . The analytical solution is

$$\mathbf{y}(t) = \begin{bmatrix} 1/(1 - \alpha \sin t) \\ \alpha \cos t/(1 - \alpha \sin t)^2 \end{bmatrix}.$$

# Problem 1: Simple Linear ODE

$$\frac{d}{dt} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} -y_1 \\ -2y_2 \\ \vdots \\ -ny_n \end{bmatrix} \Longrightarrow \mathbf{y}(t) = \begin{bmatrix} \exp(-x) \\ \exp(-2x) \\ \vdots \\ \exp(-nx) \end{bmatrix}$$
$$\mathbf{y}(0) = \begin{bmatrix} 1 \ 1 \ \cdots \ 1 \end{bmatrix}^T, t \in [0, 1/4], n = 2048.$$

# Problem 1: Simple Linear ODE

Romberg sequence: $L=4$ at $t_{ m end}=1/4$					
L = 4	Computational time (s) on Ryzen				
#steps	DD	DEFT	DEFTA	Double	DMøller
512	1.79	1.41	0.99	0.2	0.33
1024	3.59	2.81	1.95	0.41	0.67
2048	7.18	5.64	3.82	0.81	1.33
4096	14.4	11.3	7.58	1.62	2.66
#steps	C	Computational time (s) on Corei7			
512	1.17	0.86	0.73	0.1	0.26
1024	2.33	1.69	1.47	0.21	0.52
2048	4.64	3.39	2.92	0.41	1.04
4096	9.34	6.75	5.87	0.82	2.07
#steps	Max. Relative Error				
512	1.8E-07	1.8E-07	1.8E-07	1.8E-07	1.8E-07
1024	1.2E-10	1.2E-10	1.2E-10	1.2E-10	1.2E-10
2048	9.3E-14	9.3E-14	9.3E-14	1.5E-13	9.4E-14
4096	8.2E-17	4.6E-16	4.6E-16	2.3E-13	4.3E-14

# Problem 1: Simple Linear ODE

Harmonic sequence: L = 6 at  $t_{end} = 1/4$ 

L = 6	Computational Time (s) on Ryzen				
#steps	DD	DEFT	DEFTA	Double	DMøller
512	1.87	1.76	1.42	0.28	0.4
1024	3.74	3.53	2.84	0.55	0.81
2048	7.48	6.93	5.58	1.11	1.62
4096	14.9	10.4	8.38	2.22	3.24
#steps	Computational Time (s) on Corei7				
512	1.4	1.04	0.89	0.1	0.26
1024	2.8	2.07	1.78	0.21	0.52
2048	5.6	4.11	3.5	0.41	1.04
4096	11.2	6.17	5.27	0.82	2.07
#steps	Max. Relative Error				
512	4.3E-10	4.3E-10	4.3E-10	4.3E-10	4.3E-10
1024	1.7E-14	2.7E-14	2.7E-14	7.1E-13	6.6E-13
2048	8.4E-19	1.3E-14	1.3E-14	9.2E-13	7.2E-13
4096	4.6E-23	5.5E-15	5.5E-15	1.0E-12	7.6E-13

#### Problem 2: Resonance problem

We pick up the following resonance problem that is necessary to control step sizes.

$$\frac{d}{dt} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} y_2 \\ -\alpha y_1^2 \sin t + 2\alpha y_1 y_2 \cos t \end{bmatrix}$$
$$\mathbf{y}(0) = \begin{bmatrix} 1 \ \alpha \end{bmatrix}^T, \ t \in [0, 37]$$

where  $\alpha=0.99999999.$  The analytical solution is

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 1/(1-\alpha\sin t) \\ \alpha\cos t/(1-\alpha\sin t)^2 \end{bmatrix}.$$

The algorithm of step size control is the same one proposed in Murofushi and Nagasaka[4], wherein the current step size is halved if the convergent condition (4) is not satisfied. The maximum stages are L = 12 for Romberg sequence and L = 18 for harmonic sequence as recommended in [4].

Problem 2: Resonance problem

Computational time and maximum relative errors					
at $t_{ m end}=37$ with Romberg seq.					
Romberg,		Ryzen	Corei7		
L = 12	#steps	Comp.Time (s)		Max.Rel.Err.	
Double	84	0.360	0.018	1.0E-01	
DEFT	100	0.514	0.401	3.7E-04	
DEFTA	100	0.507	0.398	3.7E-04	
DMøller	98	0.149	0.094	5.2E-04	
DD	213	1.895	1.598	1.1E-17	

Problem 2: Resonance problem

at $t_{end} = 37$ with harmonic seq.						
Harmonic,		Ryzen	Corei7			
L = 18	#steps	Comp.Time (s)		Max.Rel.Err.		
Double	NC					
DEFT	159	0.0458	0.0383	4.5E-04		
DEFTA	159	0.0448	0.0378	4.5E-04		
DMøller	NC					
$DD(\varepsilon_R = 10^{-16})$	121	0.136	0.077	3.2E-02		
$DD(\varepsilon_R = 10^{-18})$	186	0.158	0.098	6.0E-05		
$DD(\varepsilon_R = 10^{-30})$	6455	1.53	1.30	4.6E-13		

Computational time and maximum relative errors

# Conclusion

- DEFTA is approximately 1.6 times faster than DD and 1.2 times faster than DEFT.
- There are no differences between DEFT's and DEFTA's approximations.
- DEFT and DEFTA are effective for resonance problem with harmonic sequence.

#### References

- T. Ogita, S. M. Rump, and S. Oishi, Accurate sum of and dot product, SIAM Journal of Scientific Computing 26(2005), 1955–1988.
- Y. Kobayashi and T. Ogita, A fast and efficient algorithm for solving ill-conditioned linear systems, JSIAM Letters **7**(2015), 1–4.
- E. Hairer, S. P. Nørsett and G. Wanner, Solving Ordinary Differential Equations I, Springer-Verlarg, New York, 1996.
- M. Murofushi and H. Nagasaka, The relationship between the round-off errors and Møller's algorithm in the extrapolation method, Annals Num., **1**(1994), 451-458.
- O. Møller, Quasi Double-Precision in Floating Point Addition, BIT 5(1965), 37-50.
- S. Bold and J. -M. Muller, Exact and Approximated Error of the FMA, IEEE Transactions on Computers, 60(2011), 157–164.