# Faithful roundings of sum with nonnegative entries

Stef Graillat ステフ・グライヤ

LIP6/PEQUAN - Université Pierre et Marie Curie (Paris 6) - CNRS

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- Computing summation is a basic task in scientific computing
- Classic algorithm is recursive summation algorithm

Algorithm 1 (Recursive summation algorithm)

```
function res = Sum(p)
s = 0
for i = 1 : n
s = fl(s + p_i)
```

res = s

• But due to rounding errors, the computed result can be far from the exact result

- Motivations
- Basic of floating-point arithmetic
- Faithful roundings of sum with nonnegative entries
- Faithful roundings of product of floating-point numbers
- Conclusion and future work

## Floating-point numbers

Normalized floating-point numbers  $\mathbb{F} \subset \mathbb{R}$ :

$$x = \pm \underbrace{x_0.x_1...x_{M-1}}_{mantissa} \times \overset{b}{b}^e, \quad 0 \le x_i \le b-1, \quad x_0 \ne 0$$

*b* : basis, *M* : precision, *e* : exponent such that  $e_{\min} \le e \le e_{\max}$ 

Approximation of  $\mathbb{R}$  by  $\mathbb{F}$  with rounding fl :  $\mathbb{R} \to \mathbb{F}$ . Let  $x \in \mathbb{R}$  then

 $fl(x) = x(1+\delta), \quad |\delta| \le \mathbf{u}$ 

Unit rounding  $\mathbf{u} = b^{1-M}/2$  for rounding to nearest

## Standard model of floating-point arithmetic

Let  $x, y \in \mathbb{F}$  and  $\circ \in \{+, -, \cdot, /\}$ .

The result  $x \circ y$  is not in general a floating-point number

 $fl(x \circ y) = (x \circ y)(1 + \delta), \quad |\delta| \le \mathbf{u}$ 

#### IEEE 754 standard (1985 and 2008)

Туре	Size	Mantissa	Exponent	Unit rounding	Interval
Single	32 bits	23+1 bits	8 bits	$\mathbf{u} = 2^{-24} \approx 5,86 \times 10^{-8}$	$\approx 10^{\pm 38}$
Double	64 bits	52+1 bits	11 bits	$\mathbf{u} = 2^{-53} \approx 1,11 \times 10^{-16}$	$\approx 10^{\pm 308}$

## Error analysis for the recursive algorithm

#### Algorithm 2 (Recursive summation algorithm)

function res = Sum(p)

s = 0for i = 1 : n $s = fl(s + p_i)$ res = s

#### Lemma 1 (Higham)

Let  $s = \sum_{i=1}^{n} p_i$ . Then we have

$$|\operatorname{res} - s| \le \gamma_{n-1} \sum_{i=1}^n |p_i|$$

where  $\gamma_n = \frac{n\mathbf{u}}{1-n\mathbf{u}}$ 

## Verified computing

As

$$|\texttt{res} - s| \le \gamma_{n-1} \sum_{i=1}^n |p_i|$$

then getting a tight error bound needs to accurately evaluate

$$\sum_{i=1}^{n} |p_i|$$

 $\rightarrow$  need to accurately evaluate the sum of nonnegative numbers

## Conditioning of summation

Condition numbers measure the sensitivity of the solution of a problem to perturbation in the data

$$\operatorname{cond}\left(\sum p_{i}\right) := \lim_{\varepsilon \to 0} \sup\left\{ \left| \frac{\sum (p_{i} + \tilde{p}_{i}) - \sum p_{i}}{\varepsilon \sum p_{i}} \right| : |\tilde{p}_{i}| \le \varepsilon |p| \right\}$$

It is well-known that

$$\operatorname{cond}\left(\sum p_i\right) = \frac{\sum |p_i|}{|\sum p_i|}$$

So for nonnegative numbers

 $\operatorname{cond}\left(\sum p_i\right) = 1$ 

The problem is then well-conditioned

## Recursive algorithm with nonnegative numbers

#### Algorithm 3 (Recursive summation algorithm)

```
function res = Sum(p)

s = 0

for i = 1 : n

s = fl(s + p_i)
```

res = s

If  $s := \sum p_i \neq 0$  then

$$\frac{|\mathsf{res} - s|}{|s|} \le \gamma_{n-1} \approx (n-1)\mathbf{u}$$

Good accuracy if *n* is small but possibly no accuracy at all if  $n \approx 1/\mathbf{u}$ 

# Getting more accuracy with compensated algorithms

Assume floating point arithmetic adhering IEEE 754 with rounding to nearest with rounding unit **u** (no underflow nor overflow)

Error free transformations are properties and algorithms to compute the generated elementary rounding errors,

*a*, *b* entries  $\in \mathbb{F}$ ,  $a \circ b = fl(a \circ b) + e$ , with  $e \in \mathbb{F}$ 

Key tools for accurate computation

- fixed length expansions libraries: double-double (Briggs, Bailey, Hida, Li), quad-double (Bailey, Hida, Li)
- arbitrary length expansions libraries: Priest, Shewchuk
- compensated algorithms (Kahan, Priest, Ogita-Rump-Oishi, Graillat-Langlois-Louvet)

## EFT for the summation

$$x = fl(a \pm b) \implies a \pm b = x + y \text{ with } y \in \mathbb{F},$$

Algorithms of Dekker (1971) and Knuth (1974)

Algorithm 4 (EFT of the sum of 2 floating point numbers with  $|a| \ge |b|$ )

function [x, y] = FastTwoSum(a, b) x = fl(a + b)y = fl((a - x) + b)

#### Algorithm 5 (EFT of the sum of 2 floating point numbers)

```
function [x, y] = TwoSum(a, b)

x = fl(a + b)

z = fl(x - a)

y = fl((a - (x - z)) + (b - z))
```

#### Theorem 1

Let  $a, b \in \mathbb{F}$  and let  $x, y \in \mathbb{F}$  such that [x, y] = TwoSum(a, b). Then,

 $a+b = x+y, \quad x = fl(a+b), \quad |y| \le \mathbf{u}|x|, \quad |y| \le \mathbf{u}|a+b|.$ 

The algorithm TwoSum requires 6 flops.

## Compensated summation algorithm



#### Algorithm 6 (Ogita, Rump, Oishi (2005))

```
function res = CompSum(p)

\pi_1 = p_1; \sigma_1 = 0;

for i = 2: n
```

$$[\pi_i, q_i] = \text{TwoSum}(\pi_{i-1}, p_i)$$
  
$$\sigma_i = \text{fl}(\sigma_{i-1} + q_i)$$

 $res = fl(\pi_n + \sigma_n)$ 

## Compensated summation algorithm

#### Algorithm 7 (Ogita, Rump, Oishi (2005))

*function* res = CompSum(*p*)

$$\begin{aligned} \pi_1 &= p_1; \sigma_1 = 0; \\ for & i = 2: n \\ & [\pi_i, q_i] = \texttt{TwoSum}(\pi_{i-1}, p_i) \\ & \sigma_i = \texttt{fl}(\sigma_{i-1} + q_i) \\ \texttt{res} &= \texttt{fl}(\pi_n + \sigma_n) \end{aligned}$$

Let  $s = \sum_{i=1}^{n} p_i$ . Then one has (Ogita, Rump, Oishi 2005)

$$|\text{res} - s| \le \mathbf{u} |\sum_{i=1}^{n} p_i| + \gamma_{n-1}^2 \sum_{i=1}^{n} |p_i|$$

where  $\gamma_n = \frac{n\mathbf{u}}{1-n\mathbf{u}}$ 

## Faithful rounding (1/3)

Floating point predecessor and successor of a real number *r* satisfying  $\min\{f : f \in \mathbb{R}\} < r < \max\{f : f \in \mathbb{F}\}$ :

pred(r) := max{ $f \in \mathbb{F} : f < r$ } and succ(r) := min{ $f \in \mathbb{F} : r < f$ }.

#### Definition 1

A floating point number  $f \in \mathbb{F}$  is called a faithful rounding of a real number  $r \in \mathbb{R}$  if

 $\operatorname{pred}(f) < r < \operatorname{succ}(f)$ .

We denote this by  $f \in \Box(r)$ . For  $r \in \mathbb{F}$ , this implies that f = r.

Faithful rounding means that the computed result is equal to the exact result if the latter is a floating point number and otherwise is one of the two adjacent floating point numbers of the exact result.

## Faithful rounding (2/3)



#### Lemma 2 (Rump, Ogita and Oishi, 2005)

Let  $r, \delta \in \mathbb{R}$  and  $\tilde{r} := \mathrm{fl}(r)$ . Suppose that  $2|\delta| < \mathbf{u}|\tilde{r}|$ . Then  $\tilde{r} \in \Box(r+\delta)$ , that means  $\tilde{r}$  is a faithful rounding of  $r+\delta$ .

## Faithful rounding (3/3)

Let res = CompSum(p)

#### Theorem 2 (Graillat 2011)

Suppose CompSum algorithm is applied to nonnegative floating-point number  $p_i \in \mathbb{F}$ ,  $1 \le i \le n$  and that

$$n < 1 + \frac{\sqrt{1-\mathbf{u}}}{\sqrt{2}\sqrt{1+\mathbf{u}} + \sqrt{1-\mathbf{u}}} \mathbf{u}^{-1/2}.$$

Then the result res is a faithful rounding of  $s := \sum p_i \ge 0$ .

If  $n < \alpha \mathbf{u}^{-1/2}$  where  $\alpha \approx 0.4$  then the result is faithfully rounded

In double precision where  $\mathbf{u} = 2^{-53}$ , if  $n \le 3 \cdot 10^7$ , we get a faithfully rounded result

S. Graillat (Univ. Paris 6)

## Classic method for computing product

The classic method for evaluating a product of *n* numbers  $a = (a_1, a_2, ..., a_n)$ 

 $p = \prod_{i=1}^{n} a_i$ 

is the following algorithm.

Algorithm 8 (Product evaluation)

```
function res = Prod(a)

p_1 = a_1

for i = 2 : n

p_i = fl(p_{i-1} \cdot a_i) \% rounding error \pi_i

end

res = p_n
```

#### This algorithm requires n-1 flops

$$\gamma_n := \frac{n\mathbf{u}}{1-n\mathbf{u}} \quad \text{for } n \in \mathbb{N}.$$

A forward error bound is

$$|a_1a_2\cdots a_n - \operatorname{res}| \le \gamma_{n-1}|a_1a_2\cdots a_n|$$

A validated error bound is

$$|a_1a_2\cdots a_n-\operatorname{res}| \le \operatorname{fl}\left(\frac{\gamma_{n-1}|\operatorname{res}|}{1-2\mathbf{u}}\right)$$

 $x = \mathrm{fl}(a \cdot b) \implies a \cdot b = x + y \text{ with } y \in \mathbb{F},$ 

Algorithm TwoProduct by Veltkamp and Dekker (1971)

a = x + y and x and y non overlapping with  $|y| \le |x|$ .

## Algorithm 9 (Error-free split of a floating point number into two parts)

function 
$$[x, y] = \text{Split}(a, b)$$
  
factor = fl(2<sup>s</sup> + 1) %  $\mathbf{u} = 2^{-p}$ ,  $s = \lceil p/2 \rceil$   
 $c = \text{fl}(\text{factor} \cdot a)$   
 $x = \text{fl}(c - (c - a))$   
 $y = \text{fl}(a - x)$ 

## EFT for the product (2/3)

## Algorithm 10 (EFT of the product of 2 floating point numbers)

 $\begin{aligned} & \text{function } [x, y] = \texttt{TwoProduct}(a, b) \\ & x = \texttt{fl}(a \cdot b) \\ & [a_1, a_2] = \texttt{Split}(a) \\ & [b_1, b_2] = \texttt{Split}(b) \\ & y = \texttt{fl}(a_2 \cdot b_2 - (((x - a_1 \cdot b_1) - a_2 \cdot b_1) - a_1 \cdot b_2))) \end{aligned}$ 

#### Theorem 3

Let  $a, b \in \mathbb{F}$  and let  $x, y \in \mathbb{F}$  such that [x, y] = TwoProduct(a, b). Then,

$$a \cdot b = x + y, \quad x = \mathrm{fl}(a \cdot b), \quad |y| \le \mathbf{u}|x|, \quad |y| \le \mathbf{u}|a \cdot b|,$$

The algorithm TwoProduct requires 17 flops.

Given  $a, b, c \in \mathbb{F}$ ,

• FMA(a, b, c) is the nearest floating point number  $a \cdot b + c \in \mathbb{F}$ 

Algorithm 11 (EFT of the product of 2 floating point numbers)

```
function [x, y] = \text{TwoProductFMA}(a, b)

x = \text{fl}(a \cdot b)

y = \text{FMA}(a, b, -x)
```

The FMA is available for example on PowerPC, Itanium, Cell processors.

## Algorithm 12 (Product evaluation with a compensated scheme (Graillat 2008))

```
function res = CompProd(a)
```

```
p_{1} = a_{1}
e_{1} = 0
for i = 2 : n
[p_{i}, \pi_{i}] = \text{TwoProduct}(p_{i-1}, a_{i})
e_{i} = \text{fl}(e_{i-1}a_{i} + \pi_{i})
end
res = fl(p_{n} + e_{n})
```

This algorithm requires 19n - 18 flops

#### Theorem 4 (Graillat 2008)

Suppose Algorithm CompProd is applied to floating point number  $a_i \in \mathbb{F}$ ,  $1 \le i \le n$ , and set  $p = \prod_{i=1}^n a_i$ . Then,

 $|\operatorname{res} - p| \le \mathbf{u}|p| + \gamma_n \gamma_{2n}|p|$ 

Condition number of the product evaluation:

$$\operatorname{cond}(a) = \lim_{\varepsilon \to 0} \sup \left\{ \frac{|(a_1 + \Delta a_1) \cdots (a_n + \Delta a_n) - a_1 \cdots a_n|}{\varepsilon |a_1 a_2 \cdots a_n|} : |\Delta a_i| \le \varepsilon |a_i| \right\}$$

A standard computation yields

cond(a) = n

#### Lemma 3 (Graillat 2008)

Suppose Algorithm CompProd is applied to floating point numbers  $a_i \in \mathbb{F}$ ,  $1 \le i \le n$  and set  $p = \prod_{i=1}^n a_i$ . Then, the absolute forward error affecting the product is bounded according to

$$|\operatorname{res} - p| \le \operatorname{fl}\left(\left(\mathbf{u}|\operatorname{res}| + \frac{\gamma_n \gamma_{2n} |a_1 a_2 \cdots a_n|}{1 - (n+3)\mathbf{u}}\right) / (1 - 2\mathbf{u})\right).$$

## Faithful rounding

Let res = CompProd(*p*)

Lemma 4  
If 
$$n < \frac{\sqrt{1-\mathbf{u}}}{\sqrt{2}\sqrt{2+\mathbf{u}}+2\sqrt{(1-\mathbf{u})\mathbf{u}}}\mathbf{u}^{-1/2}$$
 then res is a faithful rounding of p.

If  $n < \alpha \mathbf{u}^{-1/2}$  where  $\alpha \approx 1/2$  then the result is faithfully rounded

In double precision where  $\mathbf{u} = 2^{-53}$ , if  $n < 2^{25} \approx 5 \cdot 10^7$ , we get a faithfully rounded result

## Validated error bound and faithful rounding

If

$$\mathrm{fl}\!\left(2\frac{\gamma_n\gamma_{2n}|a_1a_2\cdots a_n|}{1-(n+3)\mathbf{u}}\right) < \mathrm{fl}(\mathbf{u}|\mathtt{res}|)$$

then we got a faitfully rounded result. This makes it possible to check *a posteriori* if the result is faithfully rounded.

### Various results

- Similar results apply for other compensated algorithm for dot product or Horner scheme with nonnegative entries
- Compensated dot product : computing  $x^T y$

#### Algorithm 13 (Ogita, Rump and Oishi 2005)

```
function res = Dot2(x, y)

[p, s] = TwoProduct(x_1, y_1)

for i = 2: n

[h, r] = TwoProduct(x_i, y_i)

[p, q] = TwoSum(p, h)

s = fl(s + (q + r))

end

res = fl(p + s)
```

## Compensated dot product

#### Algorithm 14 (Ogita, Rump and Oishi 2005)

```
function res = Dot2(x, y)

[p, s] = TwoProduct(x_1, y_1)

for i = 2 : n

[h, r] = TwoProduct(x_i, y_i)

[p, q] = TwoSum(p, h)

s = fl(s + (q + r))

end

res = fl(p + s)
```

Then one has (Ogita, Rump, Oishi 2005)

$$|\operatorname{res} - x^{T}y| \le \mathbf{u}|x^{T}y| + \gamma_{2n}^{2}|x|^{T}|y|$$

where  $\gamma_n = \frac{n\mathbf{u}}{1-n\mathbf{u}}$ 

## The Horner scheme

Evaluation of 
$$p(x) = \sum_{i=0}^{n} a_i x^i$$

#### Algorithm 15 (Horner scheme)

```
function res = Horner(p, x)
```

$$s_n = a_n$$
  
for  $i = n - 1 : -1 : 0$   
$$p_i = fl(s_{i+1} \cdot x)$$
  
$$s_i = fl(p_i + a_i)$$
  
ord

% rounding error  $\pi_i$ % rounding error  $\sigma_i$ 

ena

 $res = s_0$ 

$$\frac{|p(x) - \operatorname{Horner}(p, x)|}{|p(x)|} \leq \underbrace{\gamma_{2n}}_{\approx 2n\mathbf{u}} \operatorname{cond}(p, x) \text{ with } \operatorname{cond}(p, x) := \frac{\sum_{i=0}^{n} |a_i| |x|^i}{|\sum_{i=0}^{n} a_i x^i|}$$

### Error-free transformation for the Horner scheme

$$p(x) = \text{Horner}(p, x) + (p_{\pi} + p_{\sigma})(x)$$

Algorithm 16 (Error-free transformation for the Horner scheme (Graillat,Louvet,Langlois 2005))

```
function [Horner(p, x), p_{\pi}, p_{\sigma}] = EFTHorner(p, x)

s_n = a_n

for i = n - 1 : -1 : 0

[p_i, \pi_i] = TwoProduct(s_{i+1}, x)

[s_i, \sigma_i] = TwoSum(p_i, a_i)

Let \pi_i be the coefficient of degree i of p_{\pi}

Let \sigma_i be the coefficient of degree i of p_{\sigma}

end

Horner(p, x) = s_0
```

## Compensated Horner scheme and its accuracy

#### Algorithm 17 (Compensated Horner scheme)

function res = CompHorner(p, x)  $[h, p_{\pi}, p_{\sigma}] = \text{EFTHorner}(p, x)$   $c = \text{Horner}(p_{\pi} + p_{\sigma}, x)$ res = fl(h + c)

#### Theorem 5 (Graillat,Louvet,Langlois 2005)

Let p be a polynomial of degree n with floating point coefficients, and x be a floating point value. Then if no underflow occurs,

$$\frac{|\operatorname{CompHorner}(p, x) - p(x)|}{|p(x)|} \le \mathbf{u} + \underbrace{\gamma_{2n}^2}_{\approx 4n^2 \mathbf{u}^2} \operatorname{cond}(p, x).$$

### Numerical experiments: testing the accuracy

Evaluation of  $p_n(x) = (x-1)^n$  for x = fl(1.333) and n = 3, ..., 42



#### Conclusion

- Compensated algorithms make it possible to accurately compute with nonnegative entries
- It makes it possible to compute some accurate error bounds

#### Future work

• Computing accurately the 2-norm of a vector

## Thank you for your attention