# Getting tight error bounds in floating-point arithmetic: illustration with complex functions, and the real $x^n$ function

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## Floating-Point Arithmetic

- too often, viewed as a set of cooking recipes;
- too many "theorems" that hold... provided no variable is very near a power of the radix, there is no underflow/overflow; or that are dangerously generalized from radix 2 to radix 10, etc.
- simple models such as the standard model

$$\circ (a \top b) = (a \top b) \cdot (1 + \delta), \quad |\delta| \le u,$$

 $(u = 2^{-p} \text{ in radix } 2, \text{ precision-}p, \text{ rounded to nearest, arithmetic})$  do not allow to catch subtle behaviors such as those in

s = a + b ; z = s - a ; r = b - z

(fast2sum) and many others.

• by the way, are these "subtle behaviors" robust?

- revisit "folklore knowledge" on FP arithmetic, and determine which properties are really true, and in what context they are true;
- build new knowledge on FP arithmetic;
- try to get optimal/asymptotically optimal/close-to-optimal error bounds;
- $\cdots$  all this in close collaboration with the formal proof folks.

## Binary Floating-Point System

#### Parameters:

 $\left\{ \begin{array}{ll} {\rm radix \ (or \ base): 2} & {\rm here;} \\ {\rm precision} & p \geq 1 \\ {\rm extremal \ exponents} & e_{\min}, e_{\max}, \end{array} \right.$ 

A finite FP number x is represented by 2 integers:

• integral significand: M,  $|M| \le 2^p - 1$ ;

• exponent e, 
$$e_{\min} \leq e \leq e_{\max}$$
. such that

$$x = M \times 2^{e+1-p}$$

with |M| largest under these constraints ( $\rightarrow |M| \ge 2^{p-1}$ , unless  $e = e_{\min}$ ). (Real) significand of x: the number  $m = M \times 2^{1-p}$ , so that  $x = m \times 2^{e}$ .

- In general, the sum, product, quotient, etc., of two FP numbers is not an FP number: it must be rounded;
- correct rounding: Rounding function ∘, and when (a⊤b) is performed, the returned value is ∘(a⊤b);
- default rounding function RN (round to nearest even):
  - (i) for all FP numbers y,  $|RN(t) t| \le |y t|$
  - (*ii*) if there are two FP numbers that satisfy (*i*), RN(t) is the one whose integral significand is even.

- a few "basic building blocks" of numerical computing:  $ab \pm cd$ , complex arithmetic,  $x^n$ ;
- "usual" error bounds:
  - prove them;
  - try to improve them;
  - discuss their possible optimality or near-optimality.
- we assume than an FMA instruction is available: computes RN(ab + c).

(FMA: first appeared in IBMP RS/6000, then PowerPC and Itanium, now specified by IEEE 754-2008)

#### Relative error due to roundings, u, and ulp notations

Let 
$$t \in \mathbb{R}, 2^{e} \leq t < 2^{e+1}$$
, with  $e \geq e_{\min}$ ;

• we have  $2^{e} \leq \mathsf{RN}(t) \leq 2^{e+1}$ , and

$$|\mathsf{RN}(t) - t| \le 2^{e-p}.$$
 (1)

 $\rightarrow$  upper bound on the relative error due to rounding *t*:

$$\left|\frac{\mathsf{RN}(t)-t}{t}\right| \le u = 2^{-p}.$$
(2)

u = 2<sup>-p</sup>: rounding unit.
 ulp(t) = 2<sup>e-p+1</sup>.

#### Relative error due to roundings, u, and ulp notations

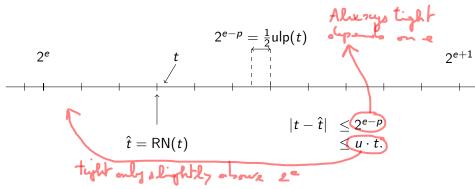


Figure 1: In precision-p binary FP arithmetic, in the normal range, the relative error due to rounding to nearest is bounded by  $u = 2^{-p}$ .

## A small improvement

The bound on the relative error due to rounding can be slightly improved (using a remark by Jeannerod and Rump):

if  $2^e \leq t < 2^{e+1}$ , then  $|t - \mathsf{RN}(t)| \leq 2^{e-p} = u \cdot 2^e$ , and

• if 
$$t \ge 2^e \cdot (1+u)$$
, then  $|t - RN(t)|/t \le u/(1+u)$   
• if  $t = 2^e \cdot (1 + \tau \cdot u)$  with  $\tau \in [0, 1)$ , then  
 $|t - RN(t)|/t = \tau \cdot u/(1 + \tau \cdot u) < u/(1+u)$ ,

 $\rightarrow$  the maximum relative error due to rounding is bounded by

 $\frac{u}{1+u}$ .

attained  $\rightarrow$  no further "general" improvement.

## "Wobbling" relative error

For  $t \neq 0$ , define

$$\overline{t} = \frac{t}{2^{\lfloor \log_2 |t| \rfloor}}.$$

#### We have,

Lemma 1

Let  $t \in \mathbb{R}$ . If  $2^e \le w \cdot 2^e \le |t| < 2^{e+1}, e \in \mathbb{Z}$ (in other words, if  $1 \le w \le |\overline{t}|$ ) then

$$\left|\frac{\mathsf{RN}(t)-t}{t}\right| \leq \frac{u}{w}.$$

(3)

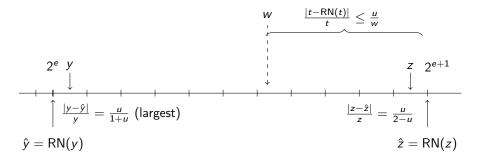


Figure 2: The bound on the relative error due to rounding to nearest can be reduced to u/(1+u). Furthermore, if we know that  $w \leq \overline{t} = t/2^e$ , then  $|\operatorname{RN}(t) - t|/t \leq u/w$ .

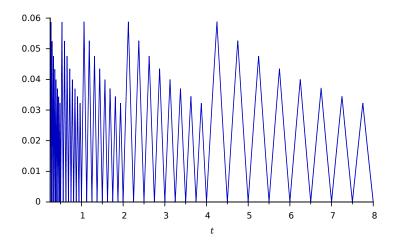


Figure 3: Relative error due to rounding, namely |RN(t) - t|/t, for  $\frac{1}{5} \le t \le 8$ , and p = 4.

#### First example: ab + cd with an FMA

Assume an fma instruction is available. Kahan's algorithm for x = ab + cd:

• using std model (Higham, 2002):

$$|\hat{x} - x| \le J|x$$

with  $J = 2u + u^2 + (u + u^2)u\frac{|cd|}{|x|} \rightarrow \text{high}$ accuracy as long as  $u|cd| \gg |x|$ 

 using properties of RN (Jeannerod, Louvet, M., 2011)

 $|\hat{x} - x| \le 2u|x|$ 

asymptotically optimal error bound.

• Complex multiplication & division.

 $\hat{w} \leftarrow \text{RN}(cd)$   $e \leftarrow \text{RN}(\hat{w} - cd)$   $\hat{f} \leftarrow \text{RN}(ab + \hat{w})$   $\hat{x} \leftarrow \text{RN}(\hat{f} - e)$ Return  $\hat{x}$ Exact quantum

#### A somewhat simpler algorithm for ab + cd

Cornea, Harrison and Tang (2002) approximate

r = ab + cd

by  $\hat{r}$  obtained as follows

algorithm CHT(a, b, c, d)  $\hat{w}_1 := RN(ab);$   $\hat{w}_2 := RN(cd);$   $e_1 := RN(ab - \hat{w}_1); e_2 := RN(cd - \hat{w}_2);$  // exact operations  $\hat{f} := RN(\hat{w}_1 + \hat{w}_2);$   $\hat{e} := RN(e_1 + e_2);$   $\hat{r} := RN(\hat{f} + \hat{e});$ return  $\hat{r};$ 

They show that the error is  $\mathcal{O}(u)$ . Since the 2u relative error bound of Kahan's algorithm was not known at that time, the CHT algorithm was favored.

We have shown the following result (ACM TOMS, to appear).

#### Theorem 2

Provided no underflow/overflow occurs, and assuming radix-2, precision-p floating-point arithmetic, the relative error of Cornea et al's algorithm is bounded by  $2u + 7u^2 + 6u^3$ .

- improvement compared to the previous  $\mathcal{O}(u)$ .
- however, does not help to choose between Kahan and CHT.

#### An almost-worst-case example...

Consider

$$\begin{cases} a = 2^{p} - 1, \\ b = 2^{p-3} + \frac{1}{2}, \\ c = 2^{p} - 1, \\ d = 2^{p-3} + \frac{1}{4}, \end{cases}$$

One easily checks that a, b, c, and d are precision-p FP numbers. One easily finds:

$$\begin{array}{rcl} ab+cd&=&2^{2p-2}+2^{p-1}-\frac{3}{4},\\ \pi_1&=&2^{2p-3}+2^{p-2},\\ e_1&=&2^{p-3}-\frac{1}{2},\\ \pi_2&=&2^{2p-3},\\ e_2&=&2^{p-3}-\frac{1}{4},\\ \pi&=&2^{2p-2},\\ e&=&2^{p-2}-\frac{3}{4},\\ s&=&2^{2p-2}. \end{array}$$

#### An almost-worst-case "generic" example...

The relative error |s - (ab + cd)|/|ab + cd| is equal to

$$\frac{2^{p-1}-\frac{3}{4}}{2^{2p-2}+2^{p-1}-\frac{3}{4}}=\frac{2u-3u^2}{1+2u-3u^2}=2u-7u^2+20u^3+\cdots$$

This shows that our relative error bound

$$2u + 7u^2 + 6u^3$$

is asymptotically optimal (as  $u \to 0$  or, equivalently, as  $p \to \infty$ ).

So that Kahan's algorithm is to be preferred, unless one wishes to get the same result when computing ab + cd and cd + ab (e.g., to get a commutative complex  $\times$ ).

Is not the theorem that gives the upper bound. It is to find the "generic" (i.e., valid  $\forall p$ ) example.

- perform the algorithm for zillions of different input values, for a given *p*, find the largest obtained relative errors,
- try to hint patterns,
- try to show that the chosen patterns effectively lead to an error close to (or, better, asymptotically equal to, or, even better, equal to) the bound.

painful, error-prone  $\rightarrow$  we are trying to (partly) automatize that step, using a "symbolic floating point" arithmetic written in Maple.

## Complex multiplication and division

Given x = a + ib and y = c + id, their product z = xy can be expressed as

z = ac - bd + i(ad + bc);

and their quotient x/y can be expressed as

$$q=\frac{ac+bd}{c^2+d^2}+i\,\frac{bc-ad}{c^2+d^2}.$$

In floating-point arithmetic, several issues:

- tradeoff accuracy vs speed,
- spurious overflow/underflow (e.g.,  $c^2 + d^2$  overflows, whereas the real and imaginary parts of q are representable);

Here: accuracy problems. Scaling techniques to avoid spurious overflow/underflow dealt with in separately. Focus on very simple algorithms.

#### Componentwise and normwise relative errors

When  $\hat{z}$  approximates z:

• componentwise error:

$$\max\left\{\left|\frac{\Re(z)-\Re(\hat{z})}{\Re(z)}\right|; \left|\frac{\Im(z)-\Im(\hat{z})}{\Im(z)}\right|\right\};$$

$$\left|\frac{z-\hat{z}}{z}\right|$$

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Choosing between both kinds of error depends on the application.

- componentwise error  $\leq \epsilon \Rightarrow$  normwise error  $\leq \epsilon$ ;
- the converse is not true.

## Naive multiplication algorithm without an FMA

$$\mathcal{A}_0$$
:  $(a+ib, c+id) \mapsto \mathsf{RN}(\mathsf{RN}(ac) - \mathsf{RN}(bd)) + i \cdot \mathsf{RN}(\mathsf{RN}(ad) + \mathsf{RN}(bc))$ 

- componentwise error: can be huge (yet finite);
- Normwise accuracy: studied by Brent, Percival, and Zimmermann (2007). The computed value has the form

$$\hat{z}_0 = z(1+\epsilon), \qquad |\epsilon| < \sqrt{5} u,$$

 $\rightarrow$  the normwise relative error  $|\hat{z}_0/z-1|$  is always  $\leq \sqrt{5} \cdot u.$ 

For any  $p \ge 2$  they provide FP numbers a, b, c, d for which  $|\hat{z}_0/z - 1| = \sqrt{5} u - O(u^2) \rightarrow$  the relative error bound  $\sqrt{5} u$  is asymptotically optimal as  $u \rightarrow 0$  (or, equivalently, as  $p \rightarrow +\infty$ ).

Can we do better if an FMA instruction is available?

## Naive multiplication algorithm with an FMA

With an FMA, the simple way of evaluating ac - bd + i(ad + bc) becomes:

$$\mathcal{A}_1 : (a + ib, c + id) \mapsto \mathsf{RN}(ac - \mathsf{RN}(bd)) + i \cdot \mathsf{RN}(ad + \mathsf{RN}(bc))$$

Algorithm  $A_1$  is just one of 4 variants that differ only in the choice of the products to which the FMA operations apply.

- componentwise error: can be huge (even infinite);
- o normwise error:
  - for any of these 4 variants the computed complex product  $\hat{z}_1$  satisfies

$$|\hat{z}_1 - z| \le 2u|z| \tag{4}$$

- we build inputs a, b, c, d for which  $|\hat{z}_1/z 1| = 2u O(u^{1.5})$  as  $u \to 0 \Rightarrow$  the error bound (4) is asymptotically optimal (given later on).
- $\rightarrow\,$  the FMA improves the situation from a normwise point of view.

## Application of Kahan's algorithm to the complex product

- $\mathbb{F}_{p}$ : precision-p, radix-2 FP numbers with unlimited exponents;
- Evaluate separately the real and imaginary parts of z = ac bd + i(ad + bc) using Kahan's algorithm;
- uses 8 floating-point operations;

$$\mathcal{A}_2: (a + ib, c + id) \mapsto \mathsf{Kahan}(a, c, -b, d) + i \cdot \mathsf{Kahan}(a, d, b, c)$$

- componentwise error  $\leq 2u$  (asymptotically optimal);
- consequence: normwise error  $\leq 2u$ .

The normwise bound is asymptotically optimal.

#### Theorem 3

Let  $a, b \in \mathbb{F}_p$  be given by

$$a = \operatorname{pred}\left(\sqrt{2^{p-2}}\right), \qquad b = 2^{p-1} + \left\lfloor\sqrt{2^{p-2}}\right\rfloor + 1,$$

where, for  $t \in \mathbb{R}_{>0}$ , pred $(t) = \max\{f \in \mathbb{F}_p : f < t\}$  denotes the predecessor of t in  $\mathbb{F}_p$ . Let also  $\hat{z}_1$  and  $\hat{z}_2$  be the approximations to  $z = (a + ib)^2$  computed by algorithms  $A_1$  and  $A_2$ , respectively. If  $p \ge 5$  then, barring underflow and overflow,

$$|\hat{z}_h/z - 1| > 2u - 8u^{1.5} - 4u^2, \qquad h \in \{1, 2\}.$$

#### Iterated products and powers

Floating-point multiplication a \* b:

- exact result z = ab;
- computed result  $\hat{z} = RN(z)$ ;

$$(1-u) \cdot z \leq \hat{z} \leq (1+u) \cdot z; \tag{5}$$

 $\rightarrow$  when we approximate  $\pi_n = a_1 \cdot a_2 \cdots \cdot a_n$  by

$$\hat{\pi}_n = \mathsf{RN}(\cdots \mathsf{RN}(\mathsf{RN}(a_1 \cdot a_2) \cdot a_3) \cdot \cdots) \cdot a_n),$$

we have

Property 1

$$(1-u)^{n-1}\pi_n \leq \hat{\pi}_n \leq (1+u)^{n-1}\pi_n.$$

#### $\gamma$ notation

 $\rightarrow$  relative error on the product  $a_1 \cdot a_2 \cdot \cdot \cdot \cdot a_n$  bounded by

$$\psi_{n-1} = (1+u)^{n-1} - 1.$$

• if we define (Higham)

$$\gamma_k = \frac{ku}{1-ku},$$

then, as long as ku < 1 (holds in practical cases),

 $\mathbf{k}\cdot\mathbf{u}\leq\psi_{\mathbf{k}}\leq\gamma_{\mathbf{k}}.$ 

 $\rightarrow$  classical relative error bound:  $\gamma_{n-1}$ .

- For "reasonable" n,  $\psi_{n-1}$  is very slightly better than  $\gamma_{n-1}$ , yet  $\gamma_{n-1}$  is easier to manipulate;
- note that in single and double precision we never observed a relative error  $\ge (n-1) \cdot u$ .

As we have seen before, the relative error bound u can be replaced by

$$\frac{u}{1+u}$$
.

 $\rightarrow$  we can replace

$$(1-u)^{n-1}\pi_n \leq \hat{\pi}_n \leq (1+u)^{n-1}\pi_n$$

by

$$\left(1-\frac{u}{1+u}\right)^{n-1}\pi_n \le \hat{\pi}_n \le \left(1+\frac{u}{1+u}\right)^{n-1}\pi_n.$$
(7)

#### Special case: $n \le 4$

Property 2

If  $1 \le k \le 3$  then

$$\left(1+\frac{u}{1+u}\right)^k < 1+k\cdot u.$$

• *k* = 2:

$$\left(1+\frac{u}{1+u}\right)^2 - (1+2u) = -\frac{u^2 \cdot (1+2u)}{(1+u)^2} < 0;$$

• *k* = 3:

$$\left(1+rac{u}{1+u}
ight)^3-(1+3u)=-rac{u^3\cdot(2+3u)}{(1+u)^3}<0.$$

 $k = n - 1 \rightarrow$  for  $n \leq 4$ , the relative error of the iterative product of n FP numbers is bounded by  $(n - 1) \cdot u$ .

## The particular case of computing powers

- "General" case of an iterated product: no proof for  $n \ge 5$  that  $(n-1) \cdot u$  is a valid bound;
- $\rightarrow$  focus on  $x^n$ , where  $x \in \mathbb{F}_p$  and  $n \in \mathbb{N}$ ;
  - we assume the "naive" algorithm is used:

```
y \leftarrow x
for k = 2 to n do
y \leftarrow RN(x \cdot y)
end for
return y
```

• notation:  $\hat{x}_j$  = value of y after the iteration corresponding to k = j in the **for** loop.

#### Main result

We are going to show:

Theorem 4

Assume  $p \ge 5$  (holds in all practical cases). If

 $n \leq \sqrt{2^{1/3}-1} \cdot 2^{p/2},$ 

then

$$|\hat{x}_n - x^n| \le (n-1) \cdot u \cdot x^n.$$

• we can assume  $1 \le x < 2$ ;

• two cases: x close to 1, and x far from 1.

## Preliminary results

First,

$$(1-u)^{n-1} \ge 1-(n-1) \cdot u$$

for all  $n \ge 2$  and  $u \in [0, 1]$ .

 $\rightarrow\,$  the left-hand bound of

$$(1-u)^{n-1}\pi_n \leq \hat{\pi}_n \leq (1+u)^{n-1}\pi_n.$$

suffices to show that

$$1-(n-1)\cdot u\cdot x_n\leq \hat{x}_n$$

 $\rightarrow\,$  to establish the Theorem, we only need to focus on the right-hand bound.

#### Reminder...

For  $t \neq 0$ , define

$$\overline{t} = \frac{t}{2^{\lfloor \log_2 |t| \rfloor}}.$$

#### We have,

Lemma 5

Let  $t \in \mathbb{R}$ . If  $2^{e} \leq w \cdot 2^{e} \leq |t| < 2^{e+1}, e \in \mathbb{Z}$ (in other words, if  $1 \leq w \leq |\overline{t}|$ ) then

$$\left|\frac{\operatorname{RN}(t)-t}{t}\right| \leq \frac{u}{w}.$$

(8)

#### Local maximum error for $x^6$ as a function of x (p = 53)

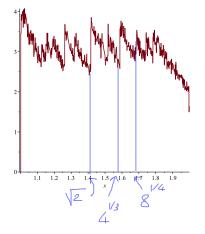


Figure 4: The input interval [1, 2) is divided into 512 equal-sized subintervals. In each subinterval, we calculate  $x^6$  for 5000 consecutive FP numbers x, compute the relative error, and plot the largest attained error.

#### Main idea behind the proof

At least once in the execution of the algorithm,  $\overline{x \cdot y}$  is far enough from 1 to sufficiently reduce the error bound on the multiplication  $y \leftarrow \text{RN}(x \cdot y)$ , so that the overall error bound becomes  $\leq (n-1) \cdot u$ .

$$y \leftarrow x$$
  
for  $k = 2$  to  $n$  do  
 $y \leftarrow RN(x \cdot y)$   
end for  
return y

$$\psi_{n-1} = (1+u)^{n-1} - 1 = (n-1)u + (1/2n^2 - 3/2n + 1)u^2 + \cdots$$

 $\rightarrow$  we have to save  $\approx \frac{n^2}{2}u^2$ , which requires one of the values  $\overline{x \cdot y}$  to be larger than  $\approx 1 + \frac{n^2}{2}u$ .

#### What we are going to show

Unless x is very near 1, at least once  $\overline{x \cdot y} \ge 1 + n^2 u$ , so that in (6) the term  $(1+u)^{n-1}$  can be replaced by

$$(1+u)^{n-2}\cdot\left(1+\frac{u}{1+n^2u}\right).$$

 $\rightarrow$  we need to bound this last quantity. We have,

Lemma 6

If  $0 \le u \le 2/(3n^2)$  and  $n \ge 3$  then

$$(1+u)^{n-2} \cdot \left(1+\frac{u}{1+n^2u}\right) \le 1+(n-1) \cdot u.$$
 (9)

Proof: tedious...

#### Two remarks

#### Remark 1

Assume  $n \leq \sqrt{2/3} \cdot 2^{p/2}$ . If  $\exists k \leq n \text{ s.t. } RN(x \cdot \hat{x}_{k-1}) \leq x \cdot \hat{x}_{k-1}$  (i.e., if in the algorithm at least one rounding is done downwards), then

 $\hat{x}_n \leq (1 + (n-1) \cdot u) x^n.$ 

#### Proof.

We have

$$\hat{x}_n \leq (1+u)^{n-2} x^n.$$

Lemma 6 implies  $(1 + u)^{n-2} < 1 + (n - 1) \cdot u$ . Therefore,

$$\hat{x}_n \leq (1+(n-1)\cdot u)x^n.$$

#### Two remarks

#### Remark 2

# Assume $n \leq \sqrt{2/3} \cdot 2^{p/2}$ . If $\exists k \leq n-1$ , s.t. $\overline{x \cdot \hat{x}_k} \geq 1 + n^2 \cdot u$ , then $\hat{x}_n \leq (1 + (n-1) \cdot u) x^n$ .

#### Proof.

By combining Lemmas 5 and 6, if there exists k,  $1 \le k \le n-1$ , such that

$$\overline{x\cdot\hat{x}_k}\geq 1+n^2\cdot u,$$

then

$$\hat{x}_n \leq (1+u)^{n-2} \cdot \left(1+rac{u}{1+n^2u}\right) \cdot x^n \leq (1+(n-1)\cdot u) \cdot x^n.$$

# Proof of Theorem 4

We assume  $n \ge 5$ . Proof articulated as follows

- if x is close enough to 1, then when computing RN(x<sup>2</sup>), the rounding is done downwards;
- in the other cases,  $\exists k \leq n-1$  such that  $\overline{x \cdot \hat{x}_k} \geq 1 + n^2 \cdot u$ .

Lemma 7

If 
$$1 < x < 1 + 2^{p/2} \cdot u$$
, then  $\hat{x}_2 = \mathsf{RN}(x^2) < x^2$ .

Proof.

 $x < 1 + 2^{p/2} \cdot u \Rightarrow x = 1 + k \cdot 2^{-p+1} = 1 + 2ku$ , with  $k < 2^{p/2-1}$ . We have  $x^2 = 1 + 2k \cdot 2^{-p+1} + k^2 \cdot 2^{-2p+2}$ , which gives  $RN(x^2) = 1 + 2k \cdot 2^{-p+1} < x^2$ .

In the following, we assume that no rounding is done downwards, which implies  $x \ge 1 + 2^{p/2} \cdot u$ .

#### Proof of Theorem 4: case $x^2 \le 1 + n^2 u$

• 
$$x \ge 1 + 2^{p/2}u > 1 + nu \Rightarrow x^n > (1 + nu)^n > 1 + n^2u;$$

• no downward rounding  $\Rightarrow \hat{x}_{n-1} \cdot x > (1 + n^2 u)$ .

Therefore

- if  $\hat{x}_{n-1}x < 2$ , then  $\overline{\hat{x}_{n-1}x} \ge (1 + n^2u)$ , so that, from Remark 2,  $x^n \le (1 + (n-1) \cdot u) \cdot x^n$ ;
- if  $\hat{x}_{n-1}x \ge 2$ , let k be the smallest integer such that  $\hat{x}_{k-1}x \ge 2$ .  $x^2 \le 1 + n^2 u \Rightarrow k \ge 3$ . We have

$$\hat{x}_{k-1} \geq \frac{2}{x} \geq \frac{2}{\sqrt{1+n^2u}},$$

hence

$$\hat{x}_{k-2} \cdot x \ge \frac{2}{\sqrt{1+n^2 u} \cdot (1+u)}.$$
 (10)

$$\hat{x}_{k-2} \cdot x \geq rac{2}{\sqrt{1+n^2u}\cdot(1+u)}$$

Define

$$\alpha_p = \sqrt{\left(\frac{2^{p+1}}{2^p+1}\right)^{2/3} - 1}.$$

For all  $p \ge 5$ ,  $\alpha_p \ge \alpha_5 = 0.745 \cdots$ , and  $\alpha_p \le \sqrt{2^{2/3} - 1} = 0.766 \cdots$ . If

$$n \le \alpha_p \cdot 2^{p/2},\tag{11}$$

•

then

$$\frac{2}{\sqrt{1+n^2u}\cdot(1+u)}\geq 1+n^2u.$$

 $\rightarrow \hat{x}_{k-2} \cdot x \ge 1 + n^2 u$ . Also,  $\hat{x}_{k-2} \cdot x < 2$  since k is the smallest integer such that  $\hat{x}_{k-1}x \ge 2$ . Therefore

$$\overline{\hat{x}_{k-2}\cdot x} \ge 1 + n^2 u.$$

Which implies  $x^n \leq (1 + (n-1) \cdot u) \cdot x^n$ .

Proof of Theorem 4: case  $x^2 > 1 + n^2 u$ 

• if 
$$x^2 < 2$$
 then  $\overline{x^2} > 1 + n^2 u \Rightarrow x^n \le (1 + (n - 1) \cdot u);$   
•  $x^2 = 2$  impossible (x is rational);  
 $\Rightarrow$  we assume  $x^2 > 2$  we also assume  $x^2 < 2 + 2n^2 u$  (otherwise,

 $\overline{x^2} \ge 1 + n^2 u$ ). This gives

$$x^{n-1} < (2+2n^2u)^{\frac{n-1}{2}},$$

therefore, using the classical bound (Property 1),

$$\hat{x}_{n-1} < (2+2n^2u)^{\frac{n-1}{2}} \cdot (1+u)^{n-2},$$

which implies

$$x \cdot \hat{x}_{n-1} < (2+2n^2u)^{\frac{n}{2}} \cdot (1+u)^{n-2}.$$
(12)

Reminder:

$$x \cdot \hat{x}_{n-1} < (2+2n^2u)^{n/2} \cdot (1+u)^{n-2}$$
 and  $n \ge 5$ 

Define

$$\beta = \sqrt{2^{1/3} - 1}.$$

If  $n \leq \beta \cdot 2^{p/2}$  then  $2 + 2n^2 u \leq 2^{4/3}$ , so that

$$(2+2n^2u)^{n/2} \cdot (1+u)^{n-2} \le 2^{2n/3} \cdot (1+u)^{n-2}.$$
(13)

The function

$$g(t) = 2^{t-1} - 2^{2t/3} \left( 1 + \frac{1}{2^p} \right)^{t-2} = 2^{2t/3} \left[ 2^{t/3-1} - \left( 1 + \frac{1}{2^p} \right)^{t-2} \right]$$

is continuous, goes to  $+\infty$  as  $t \to +\infty$ , has one root only:

$$\frac{\log(2)+2\log\left(1+\frac{1}{2^p}\right)}{\frac{1}{3}\log(2)-\log\left(1+\frac{1}{2^p}\right)},$$

which is < 4 as soon as  $p \ge 5 \Rightarrow$  if  $p \ge 5$  then  $x \cdot \hat{x}_{n-1} < 2^{n-1}$ .

Reminder: if  $p \ge 5$  then  $x \cdot \hat{x}_{n-1} < 2^{n-1}$ .

- define k as the smallest integer for which  $x \cdot \hat{x}_{k-1} < 2^{k-1}$ ,
- $3 \le k \le n$  (we have assumed  $x^2 > 2$ ),

•  $x \cdot \hat{x}_{k-2} \ge 2^{k-2} \Rightarrow \hat{x}_{k-1} = \mathsf{RN}(x \cdot \hat{x}_{k-2}) \ge 2^{k-2}.$ 

Therefore,  $\hat{x}_{k-1}$  and  $x \cdot \hat{x}_{k-1}$  belong to the same binade, therefore,

$$\overline{x \cdot \hat{x}_{k-1}} \ge x > \sqrt{2}. \tag{14}$$

The constraint  $n \leq \beta \cdot 2^{p/2}$  implies

$$1 + n^2 u \le 1 + \beta^2 = 2^{1/3} < \sqrt{2}.$$
 (15)

By combining (14) and (15) we obtain

$$\overline{x\cdot \hat{x}_{k-1}} \ge 1 + n^2 u.$$

Therefore, using Remark 2, we deduce that  $\hat{x}_n \leq (1 + (n - 1) \cdot u) \cdot x^n$ .

## Final steps

 $\forall p \geq 5, \ \alpha_p \geq \beta \rightarrow \text{ combining the conditions found in the cases}$  $x^2 \leq 1 + n^2 u$  and  $x^2 > 1 + n^2 u$ , we deduce

If 
$$p \ge 5$$
 and  $n \le \beta \cdot 2^{p/2}$ , then for all  $x$ ,  
 $(1 - (n - 1) \cdot u) \cdot x^n \le \hat{x}_n \le (1 + (n - 1) \cdot u) \cdot x^n$ .  
where  $\beta = \sqrt{2^{1/3} - 1} = 0.5098245285339 \cdots$ 

Q.E.D.

Questions:

- is the restriction  $n \leq \beta \cdot 2^{p/2}$  problematic?
- is the bound sharp?

format	р	n <sub>max</sub>
binary32/single	24	2088
binary64/double	53	48385542
binary128/quad	113	51953580258461959

With the first *n* larger than the bound,  $x^n$  under- or overflows, unless

- in single precision,  $0.95905406 \le x \le 1.0433863$ ,
- in double precision,  $0.999985359 \le x \le 1.000014669422$ ,

and nobody will use the "naive" algorithm for a huge n.

Furthermore, that restriction is not just a "proof artefact". For very big n, the bound does not hold:

If p = 10 and x = 891, when computing  $x^{2474}$ , relative error 2473.299u.

Notice that:

• for p = 10,  $n_{\max} = \beta \cdot 2^{p/2} = 16.31$ ;

• 2474 is the smallest exponent for which the bound does not hold when p = 10.

## Tightness of the bound $(n-1) \cdot u$

Small p and not-too-large n: an exhaustive test is possible.

Table 1: Actual maximum relative error assuming p = 8, compared with  $\gamma_{n-1}$  and our bound (n-1)u.

n	actual maximum	$\gamma_{n-1}$	our bound
4	1.73903 <i>u</i>	3.0355 <i>u</i>	3 <i>u</i>
5	2.21152 <i>u</i>	4.06349 <i>u</i>	4 <i>u</i>
6	2.53023 <i>u</i>	5.099601 <i>u</i>	5 <i>u</i>
7	2.69634 <i>u</i>	6.1440 <i>u</i>	6 <i>u</i>
$8 = n_{max}$	3.42929 <i>u</i>	7.1967 <i>u</i>	7 <i>u</i>

 $\rightarrow$  our bound seems to be quite poor... however...

For larger values of p:

- single precision (p = 24), exhaustive search still possible, largest error 4.328005619u for n = 6, and 7.059603149u for n = 10;
- double precision (p = 53), we have a case with error 4.7805779u for n = 6 and 7.8618 $\cdots u$  for n = 10;
- quad precision (p = 113), case with error 4.8827888185u for n = 6;
- $\rightarrow$  we seem to get close to  $(n-1) \cdot u$  for large p.

#### Building "bad cases" for the iterated product

Still in precision-p binary FP arithmetic, we approximate

 $a_1 \cdot a_2 \cdots \cdot a_n$ ,

by

 $RN(\cdots RN(RN(a_1 \cdot a_2) \cdot a_3) \cdot \cdots) \cdot a_n)$ 

• 
$$\pi_k = a_1 \cdots a_k$$
,

- $\hat{\pi}_k = \text{computed value},$
- relative error  $|\pi_n \hat{\pi}_n|/\pi_n$  upper-bounded by  $\gamma_{n-1}$ ,
- conjecture: if n is "not too large" it is bounded by (n-1)u. Let us now show how to build  $a_1, a_2, \ldots, a_n$  so that the relative error

becomes extremely close to  $(n-1) \cdot u$ .

#### Building "bad cases" for the iterated product

• define 
$$a_1 = 1 + k_1 \cdot 2^{-p+1}$$
, and  $a_2 = 1 + k_2 \cdot 2^{-p+1}$ . We have  
 $\pi_2 = a_1 a_2 = 1 + (k_1 + k_2) \cdot 2^{-p+1} + k_1 k_2 \cdot 2^{-2p+2}$ .

If  $k_1$  and  $k_2$  are not too large,  $1 + (k_1 + k_2) \cdot 2^{-p+1}$  is a FP number  $\rightarrow$  we wish  $k_1 + k_2$  to be as small as possible, while  $k_1 k_2 \cdot 2^{-2p+2}$  is as close as possible (yet less than) to  $2^{-p}$ . Hence a natural choice is

$$k_1=k_2=\left\lfloor 2^{\frac{p}{2}-1}\right\rfloor,$$

which gives  $\hat{\pi}_2 < \pi_2$ .

• Now, if at step i - 1 we have

$$\hat{\pi}_i = 1 + g_i \cdot 2^{-p+1}$$
, with  $\hat{\pi}_i < \pi_i$ ,

we choose  $a_{i+1}$  of the form  $1 + k_{i+1}2^{-p+1}$ , with

• 
$$k_{i+1} = \left\lceil \frac{2^{p-2}}{g_i} - 1 \right\rceil$$
 if  $g_i \le 2^{\frac{p}{2}-1}$ ;  
•  $k_{i+1} = -\left\lfloor \frac{2^{p-2}}{g_i} + 1 \right\rfloor$  otherwise.

#### Building "bad cases" for the iterated product

Table 2: Relative errors achieved with the values  $a_i$  generated by our method.

р	n	relative error
24	10	8.99336984 · · · <i>u</i>
24	100	98.9371972591 · · · <i>u</i>
53	10	8.99999972447 · · · <i>u</i>
53	100	98.9999970091 · · · <i>u</i>
113	10	8.999999999999999973119··· <i>u</i>
113	100	98.999999999999999701662··· <i>u</i>

## Conclusion on $x^n$

- error bound  $(n-1) \cdot u$  for computation of  $x^n$  by the naive algorithm;
- valid for  $n \leq \sqrt{2^{1/3} 1} \cdot 2^{p/2} \rightarrow$  all practical cases;
- small improvement: the main interest lies in the simplicity of the bound;
- ullet seems to be "asymptotically sharp" (as  $p 
  ightarrow \infty$ ) but not sure;
- the bound  $\gamma_{n-1}$  on iterated products is very sharp.

Thank you for your attention.