

Rounding Error Analysis of an Orbital Collision Probability Evaluation Algorithm

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Denis Arzelier^{*}, **Florent Bréhard[†]**, **Mioara Joldes^{*}**, **Marc Mezzarobba[°]**

^{*} LAAS–CNRS, Toulouse, France

[†] CNRS, Univ. Lille, CRISAL, France

[°] CNRS, LIX, Palaiseau, France

✉ arzelier@laas.fr; florent.brehard@univ-lille.fr; mmjoldes@laas.fr;
marc@mezzarobba.net

Recurrences and Validated Numerics

- Recurrences are ubiquitous in computer-aided mathematics: spectral methods for functional equations, dynamical systems, combinatorics, asymptotics, ...
- They were extensively studied in numerical analysis: Olver, Miller, Wimp, Clenshaw, Henrici, Barrio, ...
- Challenges for validated numerics: numerical instability, wrapping effect
⇒ Tight bounds are difficult to obtain
- M. Mezzarobba. Rounding Error Analysis of Linear Recurrences Using Generating Series. *ETNA - Electronic Transactions on Numerical Analysis*, 2023
⇒ Connect the total error to the analytic properties of the initial problem
⇒ Deduce a priori and realistic error bounds

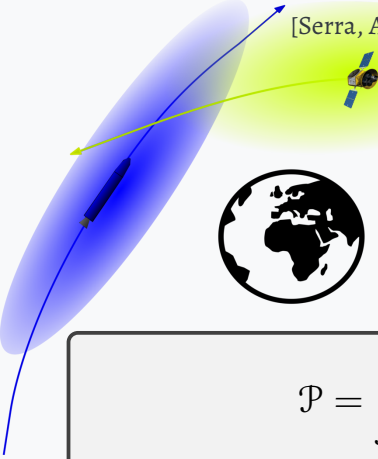
1. Orbital Collision Probability Evaluation Algorithm

2. Roundoff Error Analysis using Majorizing Series

3. Numerical Examples

Orbital Collision Probability

[Serra, Arzelier, Joldes, Lasserre, Rondepierre, Salvy – 2015]

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- The diagram illustrates the orbital collision probability (OCP) concept. It shows a satellite in a circular orbit around Earth, represented by a globe. A blue elliptical region represents the satellite's field of view or a specific orbital region. A yellow elliptical region represents the collision volume or a specific orbital region. A green arrow points from the satellite towards the collision volume, and a blue arrow points from the collision volume towards the satellite. The collision volume is defined by a green ellipse centered on the satellite's position.
- Assumptions for Short Term Encounter
 - Extensively tested and approved by CNES
 - implemented for both ground and onboard usage

$$\mathcal{P} = \iint_{x^2+y^2 \leq R^2} \rho(x,y) dx dy$$

$$\rho(x,y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp \left[-\frac{1}{2} \left(\frac{(x-x_m)^2}{\sigma_x^2} + \frac{(y-y_m)^2}{\sigma_y^2} \right) \right]$$

Algorithm: Computation of the Probability of Collision

Require: FP Parameters $R, \sigma_x, \sigma_y, x_m, y_m$; number of terms N .

Ensure: $\mathcal{P}_{0:N}$ – truncated series approximation of $\mathcal{P}$.

1: Evaluate $p, \phi, \omega_x, \omega_y, Q_1, Q_2, Q_3, P_0, P_1, P_2, P_3$;

2: $c_0 = \dots; c_1 = \dots; c_2 = \dots; c_3 = \dots$;

3: $s = c_0 + c_1 + c_2 + c_3$;

4: **for** $n = 4$ to $N - 1$ **do**

$$\begin{aligned} 5: \quad c_n = & \frac{Q_1(n-1) + P_0}{(n+1)n} c_{n-1} - \frac{Q_2(n-2) + P_1}{(n+1)n^2} c_{n-2} \\ & + \frac{Q_3(n-3) + P_2}{(n+1)n^2(n-1)} c_{n-3} - \frac{P_3}{(n+1)n^2(n-1)(n-2)} c_{n-4}; \end{aligned}$$

6: $s = s + c_n$;

7: **end for**

8: **return** $\mathcal{P}_{0:N} = \exp(-pR^2)s$.

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preconditioning
to ensure $c_n \geq 0$

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explicit truncation
error bounds

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to ensure $c_n \geq 0$

initial terms

unroll the recurrence

sum the terms

How this algorithm works in a nutshell

$$\mathcal{P} = \exp(-pR^2\xi)f(\xi) \text{ at } \xi = 1$$

$$f(\xi) = c_0 + c_1\xi + c_2\xi^2 + \dots + c_n\xi^n + \dots$$

$\Rightarrow$ ensures $c_n \geq 0$

differential equation on $f(\xi)$

D-finite functions

Recurrence on the coefficients c_n :

$$c_n = \square c_{n-1} + \square c_{n-2} + \square c_{n-3} + \square c_{n-4}$$

$$\mathcal{P}_{0:N} \stackrel{\text{def}}{=} \exp(-pR^2) \sum_{n=1}^{N-1} c_n \approx \mathcal{P}$$

How this algorithm works in a nutshell

Laplace transform

$$\mathcal{P} = \exp(-pR^2\xi)f(\xi) \text{ at } \xi = 1$$
$$f(\xi) = c_0 + c_1\xi + c_2\xi^2 + \cdots + c_n\xi^n + \cdots$$

$$\widehat{f}(\lambda) = \left\{ \text{explicit function of } \lambda \right\}$$
$$= c_0 + c_1\lambda + 2c_2\lambda^2 + \cdots + n!c_n\lambda^n + \cdots$$

 $\Rightarrow \text{ensures } c_n \geq 0$ differential equation on $f(\xi)$

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Roundoff error bound for $|\tilde{\mathcal{P}}_{0:N} - \mathcal{P}_{0:N}|$?

- standard binary64 (no interval arithmetic, no multiprecision)
 $\rightsquigarrow$ efficiency/architecture constraints
 - N can be large! ($N = 100, 1000, 10000, \dots$)
- $\Rightarrow$ A priori bounds using majorizing series techniques for round-off analysis of linear recurrences

$$c_n = \frac{Q_1(n-1) + P_0}{(n+1)n} c_{n-1} - \frac{Q_2(n-2) + P_1}{(n+1)n^2} c_{n-2} \\ + \frac{Q_3(n-3) + P_2}{(n+1)n^2(n-1)} c_{n-3} - \frac{P_3}{(n+1)n^2(n-1)(n-2)} c_{n-4}$$

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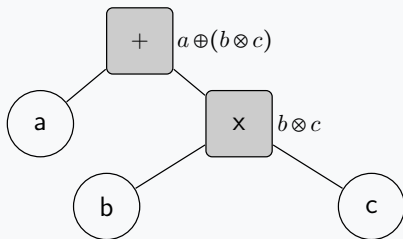
Error model for floating-point arithmetic

- Precision p binary floating-point arithmetic (e.g., $p = 53$ for binary64)
- unbounded exponent range ($\Rightarrow$ no underflow, no overflow)
- round-to-nearest mode for arithmetic operations $\star \in \{+, -, \times, \div, \sqrt{\cdot}\}$

$$\widetilde{a \star b} = (a \star b)(1 + e) \qquad |e| \leq u \stackrel{\text{def}}{=} 2^{-p}$$

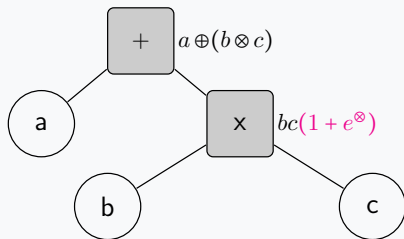
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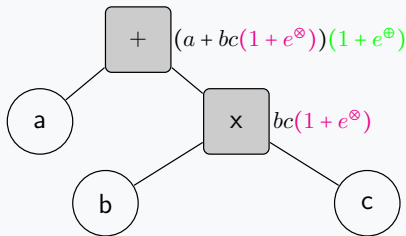
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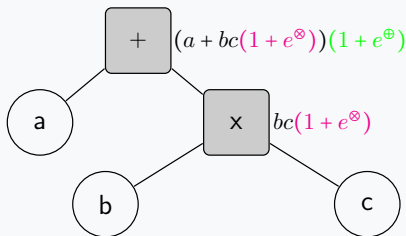
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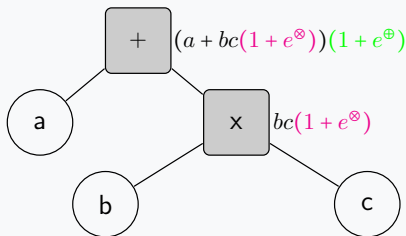


Example

$$\begin{aligned}(a \oplus b \otimes c) - (a + bc) &= \\ a e^+ + bc[(1 + e^+)(1 + e^\otimes) - 1] \\ |e^+|, |e^\otimes| &\leq u \stackrel{\text{def}}{=} 2^{-p}\end{aligned}$$

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Notation ([Higham, *Accuracy and Stability of Numerical Algorithms* – 2002])

$$(1 + e_1) \dots (1 + e_n) \stackrel{\text{def}}{=} (1 + \theta_n)$$

$$|\theta_n| \leq \gamma_n \stackrel{\text{def}}{=} \frac{nu}{1 - nu}$$

Roundoff errors in the algorithm

Require: FP Parameters $R, \sigma_x, \sigma_y, x_m, y_m$; number of terms N .

Ensure: $\mathcal{P}_{0:N}$ – truncated series approximation of $\mathcal{P}$.

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Initial errors

Local errors

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Initial errors

Local errors → Global errors

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Initial errors

Local errors $\rightarrow$ Global errors

Summation errors

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3: $s = c_0 + c_1 + c_2 + c_3$;

4: **for** $n = 4$ to $N - 1$; **do**

5:
$$c_n = \frac{Q_1(n-1)+P_0}{(n+1)n}c_{n-1} - \frac{Q_2(n-2)+P_1}{(n+1)n^2}c_{n-2}$$

$$+ \frac{Q_3(n-3)+P_2}{(n+1)n^2(n-1)}c_{n-3} - \frac{P_3}{(n+1)n^2(n-1)(n-2)}c_{n-4};$$

6: $s = s + c_n$;

7: **end for**

8: **return** $\mathcal{P}_{0:N} = \exp(-pR^2)s$.

Initial errors + Local errors $\rightarrow$ Global errors + Summation errors

Initial and local errors using standard roundoff analysis

- Initial errors (loop-independent parameters):

$$\omega_x = \frac{x_m^2}{4\sigma_x^4} \rightsquigarrow \tilde{\omega}_x = \frac{1}{4} \cdot (x_m \otimes x_m) \oslash ((\sigma_x \otimes \sigma_x) \otimes (\sigma_x \otimes \sigma_x))$$

$$\Rightarrow |\tilde{\omega}_x - \omega_x| \leq \gamma_5 \omega_x$$

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$\rightsquigarrow$ Similar bounds for the other parameters $\rightsquigarrow P_i, Q_i$

- Local errors ε_n (at each iteration):

$$\begin{aligned} \tilde{c}_n &= \frac{Q_1(n-1) + P_0}{(n+1)n} \tilde{c}_{n-1} - \frac{Q_2(n-2) + P_1}{(n+1)n^2} \tilde{c}_{n-2} \\ &+ \frac{Q_3(n-3) + P_2}{(n+1)n^2(n-1)} \tilde{c}_{n-3} - \frac{P_3}{(n+1)n^2(n-1)(n-2)} \tilde{c}_{n-4} + \varepsilon_n \end{aligned}$$

$$\begin{aligned} |\varepsilon_n| \leq \gamma_{40} &\left(\frac{Q_1(n-1) + P_0}{(n+1)n} |\tilde{c}_{n-1}| + \frac{Q_2(n-2) + P_1}{(n+1)n^2} |\tilde{c}_{n-2}| \right. \\ &\left. + \frac{Q_3(n-3) + P_2}{(n+1)n^2(n-1)} |\tilde{c}_{n-3}| + \frac{P_3}{(n+1)n^2(n-1)(n-2)} |\tilde{c}_{n-4}| \right) \end{aligned}$$

Global error: Outline

following [Mezzarobba 2020]

$$f(\xi) = \sum_{n=0}^{+\infty} c_n \xi^n$$

$$\hat{f}(\lambda) = \sum_{n=0}^{+\infty} n! c_n \lambda^n$$

$$\hat{f}'(\lambda) - \varphi(\lambda) \hat{f}(\lambda) = 0$$

$$c_n - \square c_{n-1} + \square c_{n-2} - \square c_{n-3} + \square c_{n-4} = 0$$

$$f(1) = s = \sum_{n=0}^{+\infty} c_n$$

$$\delta_n \stackrel{\text{def}}{=} \tilde{c}_n - c_n$$

$$\delta(\xi) = \tilde{f}(\xi) - f(\xi) = \sum_{n=0}^{+\infty} \underbrace{(\tilde{c}_n - c_n)}_{\delta_n} \xi^n$$

$$[\hat{\delta}(\lambda) \rightsquigarrow \delta(\xi)] \quad \uparrow \quad [\text{solve for } \delta(\lambda)]$$

$$\hat{\delta}'(\lambda) - \varphi(\lambda) \hat{\delta}(\lambda) = \hat{\varepsilon}(\lambda)$$

$$\uparrow \quad \left[\sum_{n=0}^{+\infty} (\cdot) n! \lambda^n \right]$$

$$\delta_n - \square \delta_{n-1} + \square \delta_{n-2} - \square \delta_{n-3} + \square \delta_{n-4} = \varepsilon_n$$

[bound ε_n]

$$\delta(1) = \tilde{s} - s = \sum_{n=0}^{+\infty} \tilde{c}_n - \sum_{n=0}^{+\infty} c_n$$

Global error: Differential inequality

Recall that for $\tilde{c}_n = c_n + \delta_n$,

$$\delta_n - \square \delta_{n-1} + \square \delta_{n-2} - \square \delta_{n-3} + \square \delta_{n-4} = \varepsilon_n$$

$$|\varepsilon_n| \leq \gamma_{40} \left(\square |\tilde{c}_{n-1}| + \square |\tilde{c}_{n-2}| + \square |\tilde{c}_{n-3}| + \square |\tilde{c}_{n-4}| \right) \quad \gamma_{40} \approx 40u$$

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Hence we obtain a differential inequality:

$$\hat{\delta}'(\lambda) - \varphi(\lambda)\hat{\delta}(\lambda) \ll \gamma_{40} \left(\psi(\lambda)\hat{f}(\lambda) + \underbrace{(\dots)\hat{\delta}'(\lambda) + (\dots)\hat{\delta}(\lambda)}_{\text{small}} \right)$$

$$[\psi(\lambda) = \{\text{an explicit rational function in } \lambda\}]$$

$$a(\lambda) \ll b(\lambda) \quad \text{means} \quad |a_n| \leq |b_n| \quad \text{for all} \quad n \geq 0$$

Global error: Linearized bound

- Differential inequality:

$$\hat{f}'(\lambda) - \varphi(\lambda)\hat{f}(\lambda) = o$$

$$\hat{\delta}'(\lambda) - \varphi(\lambda)\hat{\delta}(\lambda) \ll 4ou \psi(\lambda)\hat{f}(\lambda) + o(u)$$

$[\psi(\lambda)$ an explicit rational function]

- Solve to obtain an upper bound (neglecting $o(u)$ terms):

$$\hat{\delta}(\lambda) \ll \left(\delta_o + 4ou \Psi(\lambda) \right) \hat{f}(\lambda) \qquad \Psi(\lambda) \stackrel{\text{def}}{=} \int_o^\lambda \psi(\sigma) d\sigma$$

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relative error on $\tilde{c}_o$

propagation of local errors

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relative error on $\tilde{c}_o$

propagation of local errors

- Inverse Laplace transform using  Maple

$$|\delta(1)| = \sum_{n=0}^{+\infty} |\tilde{c}_n - c_n| \leq \left(\delta_o + 4ouC \right) f(1) + o(u)$$

$$C \stackrel{\text{def}}{=} \frac{7}{96} p^3 \omega_x R^8 + \left(\frac{7}{12} p + \frac{1}{2} \omega_x \right) p^2 R^6 + \left(\frac{9}{4} p + \frac{5}{4} \omega_x + \frac{15}{4} \omega_y \right) p R^4 + \left(\frac{3}{2} p + \omega_x + 3 \omega_y \right) R^2$$

Total roundoff error

Linearized bound

$$\frac{|\tilde{\mathcal{P}}_{0:N} - \mathcal{P}_{0:N}|}{\mathcal{P}} \leq \left(\frac{2x_m^2}{\sigma_x^2} + \frac{2y_m^2}{\sigma_y^2} + 40C + N + 2pR^2 + 8 \right) u + o(u)$$

$$C \stackrel{\text{def}}{=} \frac{7}{96} p^3 \omega_x R^8 + \left(\frac{7}{12} p + \frac{1}{2} \omega_x \right) p^2 R^6 + \left(\frac{9}{4} p + \frac{5}{4} \omega_x + \frac{15}{4} \omega_y \right) p R^4 + \left(\frac{3}{2} p + \omega_x + 3 \omega_y \right) R^2$$

[initial error on $\tilde{c}_0$ global error on $\tilde{c}_n$ summation error final rescaling]

- The required number of terms N depends on the parameters ($\rightarrow$ truncation error bound), and $N \leq C$ in practice
- We also derived a rigorous (i.e., not linearized) total error bound
- We proved that the 64-bit exponent emulation is sufficient to avoid overflows for realistic parameter ranges

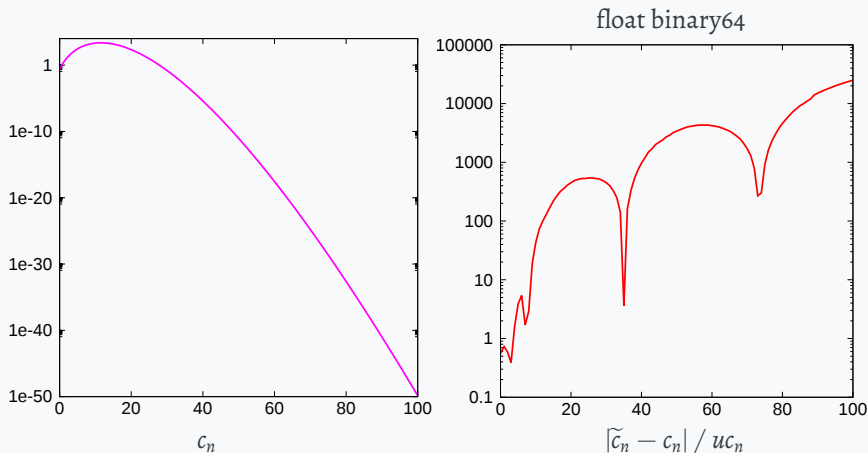
1. Orbital Collision Probability Evaluation Algorithm

2. Roundoff Error Analysis using Majorizing Series

3. Numerical Examples

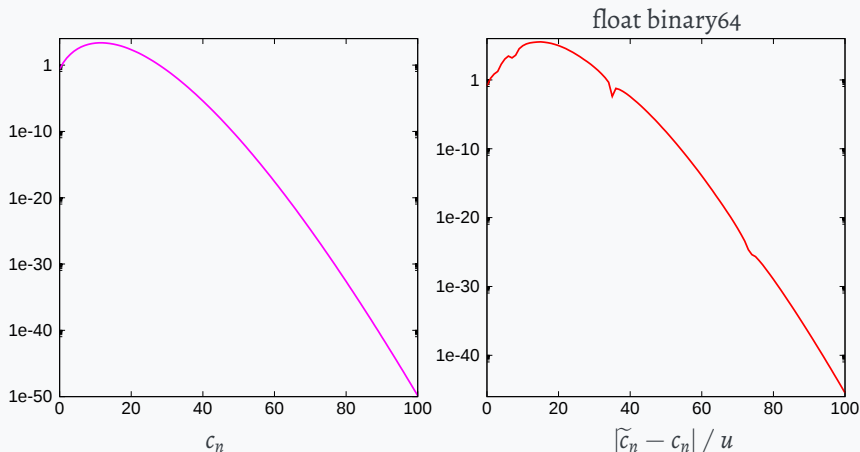
Analysis of a numerical example

$$R = 5, \quad \sigma_x = 50, \quad \sigma_y = 1, \quad x_m = 10, \quad y_m = 0 \quad \Rightarrow \quad N = 101$$



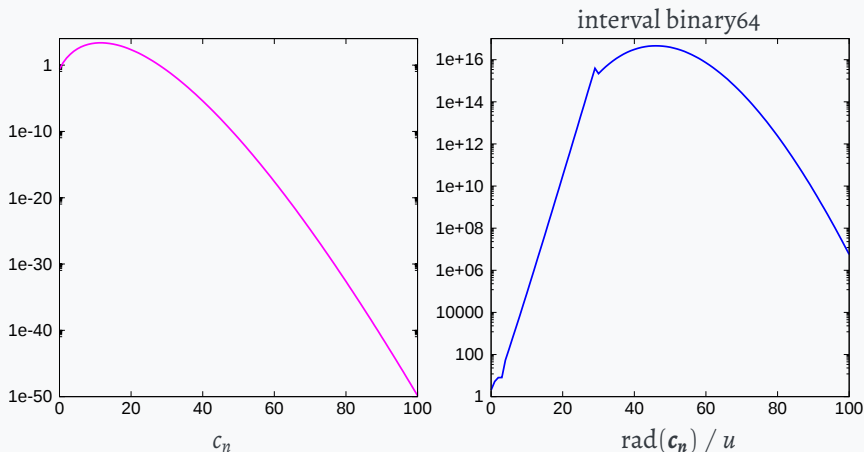
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float binary64	interval binary64	our bound
$1.26 \cdot 10^2 u$ $= 1.40 \cdot 10^{-14}$	$3.85 \cdot 10^{13} u$ $= 4.27 \cdot 10^{-3}$	$6.05 \cdot 10^4 u$ $= 6.72 \cdot 10^{-12}$

Obtained relative errors

Some more numerical examples

Case #	Input parameters (m)						Relative Error			
	σ_x	σ_y	R	x_m	y_m	N	Exact	MPFI	(Lin. Bound)	Full Bound
Test 1	50	1	5	10	0	101	1.40e-14	4.27e-3	6.72e-12	6.72e-12
Chan 1	50	25	5	10	0	49	5.86e-17	5.86e-15	6.48e-15	6.48e-15
Chan 5	3,000	1,000	10	1,000	0	49	2.02e-16	7.41e-15	6.35e-15	6.35e-15
Chan 6	3,000	1,000	10	0	1,000	48	1.18e-16	5.61e-15	6.44e-15	6.44e-15
Alfano 3	114.25	1.41	15	0.15	-3.88	1627	4.14e-12	1.15e54	7.07e-10	7.08e-10
Alfano 5	177.81	0.03	10	2.12	-1.22	> 1e7	4.35e-4	4e69380	4.87e-01	3.60e+00
Custom 1	1	1	10	1	1	543	6.96e-16	1.78e-13	1.53e-09	1.53e-09
Custom 2	1	0.8	10	1	1	969	2.73e-14	4.7e23	5.59e-09	5.60e-09
Custom 3	1	0.5	10	1	1	3805	7.74e-14	4.4e174	8.95e-08	9.00e-08
Custom 4	1	0.2	10	1	1	95139	4.6e-12	2e1483	2.13e-05	2.22e-05
Custom 5	1	0.1	10	1	1	> 1e7	3.63e-8	1e6155	1.36e-03	1.59e-03
Custom 6	0.5	0.1	10	1	1	> 1e7	1.49e-11	2e5988	1.66e-02	1.95e-02
Custom 7	1	0.05	10	1	1	> 1e7	3.00e-6	4e24841	8.68e-02	1.70e-01
Custom 8	0.2	0.05	10	1	1	> 1e7	1.28e-9	2e23506	4.05e+01	7.40e+17

Take Away Message

- Successful application of Mezzarobba's FP error analysis to this POC algorithm $\Rightarrow$ a priori closed-form relative error bounds
- Provides rigorous and realistic error bounds that could not be obtained using plain interval arithmetic
- An interesting alternative to fixed-point based a posteriori validation: the rigorous computation part is “constant time” (i.e., not proportional to N)
- However, preliminary pen-and-paper work is still substantial, the class of treated recurrences is limited (linear?), and error bounds are “worst-case”.