

# Computing the Frobenius norm using `cr_hypot`

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Consider a well-defined `hypot` function, i.e.,

$$\text{hypot}(x, y) = \sqrt{x^2 + y^2}(1 + \epsilon'),$$

where  $|\epsilon'| \ll 1$ . If it is correctly rounded for all inputs, denote it by `cr_hypot`.

**Theorem 1.** *Let  $\mathbf{x} = [x_1 \ \cdots \ x_n]^T$  be a vector of finite floating-point values, and  $\|\mathbf{x}\|_F$  its Frobenius norm. If its approximation  $\underline{\|\mathbf{x}\|_F}$  is computed as  $\underline{\|\mathbf{x}\|_F} = \underline{f}_n$ , where  $\underline{f}_0 = f_0 = 0$ ,  $\underline{f}_1 = f_1 = |x_1|$ , and*

$$2 \leq i \leq n \implies \underline{f}_i = \text{hypot}(\underline{f}_{i-1}, x_i), \quad (1)$$

*then, barring any overflow and inexact underflow, when  $x_i \neq 0$  it holds*

$$\underline{f}_i = f_i(1 + \epsilon_i), \quad 1 + \epsilon_i = \sqrt{1 + \epsilon_{i-1}(2 + \epsilon_{i-1}) \frac{f_{i-1}^2}{f_i^2}}(1 + \epsilon'_i). \quad (2)$$

*Assume that `hypot` is `cr_hypot`. Then,  $|\epsilon'_i| \leq \varepsilon$ , with  $\varepsilon$  being the machine precision. If  $x_i = 0$ , then  $\underline{f}_i = \underline{f}_{i-1}$ . If a lower bound of  $\epsilon_{i-1}$  is denoted by  $\epsilon_{i-1}^-$  and an upper bound by  $\epsilon_{i-1}^+$ , where  $\epsilon_1^- = \epsilon_1^+ = 0$ , then, when  $0 \geq \epsilon_{i-1}^- \geq -1$ , the relative error factor  $1 + \epsilon_i$  from (2) can be bounded irrespectively of  $x_i \neq 0$  and  $f_i$  as*

$$\begin{aligned} 1 + \epsilon_i^- &= \sqrt{1 + \epsilon_{i-1}^-(2 + \epsilon_{i-1}^-)}(1 - \varepsilon) \leq 1 + \epsilon_i \\ &\leq \sqrt{1 + \epsilon_{i-1}^+(2 + \epsilon_{i-1}^+)}(1 + \varepsilon) = 1 + \epsilon_i^+. \end{aligned} \quad (3)$$

*Proof.* Using any well-defined `hypot` function, assuming that (2) holds for all  $j$  such that  $0 \leq j < i$ , where  $2 \leq i \leq n$ , and that  $x_i \neq 0$ , it follows

$$\underline{f}_i = \sqrt{\underline{f}_{i-1}^2 + x_i^2}(1 + \epsilon'_i) = \sqrt{f_{i-1}^2(1 + \epsilon_{i-1})^2 + x_i^2}(1 + \epsilon'_i). \quad (4)$$

If the term under the square root on the right hand side of (4) is written as

$$f_{i-1}^2(1 + \epsilon_{i-1})^2 + x_i^2 = (f_{i-1}^2 + x_i^2)(1 + y), \quad (5)$$

then an easy algebraic manipulation gives

$$y = \epsilon_{i-1}(2 + \epsilon_{i-1}) \frac{f_{i-1}^2}{f_{i-1}^2 + x_i^2} = \epsilon_{i-1}(2 + \epsilon_{i-1}) \frac{f_{i-1}^2}{f_i^2}, \quad (6)$$

what, after taking the absolute value of both sides of (6), leads to

$$|y| \leq |\epsilon_{i-1}|(2 + |\epsilon_{i-1}|) \frac{f_{i-1}^2}{f_i^2} \leq |\epsilon_{i-1}|(2 + |\epsilon_{i-1}|). \quad (7)$$

Substituting (5) into (4) yields

$$\underline{f}_i = \sqrt{f_{i-1}^2 + x_i^2} \sqrt{1 + y} (1 + \epsilon'_i) = f_i \sqrt{1 + y} (1 + \epsilon'_i) = f_i (1 + \epsilon_i),$$

where  $(1 + \epsilon_i) = \sqrt{1 + y} (1 + \epsilon'_i)$ , as claimed in (2). The recurrence (3) for bounds on  $1 + \epsilon_i$  when `hypot` is `cr_hypot` follows from (7) and the fact that the function  $x \mapsto x(2 + x)$  is monotonically increasing for  $x \geq -1$ .  $\square$

More practically, (3) can be further simplified as  $-\epsilon_i^- < \epsilon_i^+ < i\varepsilon$  when

$$((\varepsilon = 2^{-24}) \wedge (3 \leq i \leq 5793)) \quad \vee \quad ((\varepsilon = 2^{-53}) \wedge (3 \leq i \leq 134217729)),$$

what follows from evaluating (3) iteratively over  $i$  using the MPFR library [1] with 2048 bits of precision.

A result similar to Theorem 1 can be obtained in the case of combining two partial norms as

$$\underline{f}_{[i,j]} = \text{hypot}(\underline{f}_{[i]}, \underline{f}_{[j]}),$$

e.g., when OpenMP-reducing the partial, per-thread results. Bear in mind that the OpenMP standard [2, §7.6.7] leaves the reduction order unspecified.

**Theorem 2.** *Assume that  $\underline{f}_{[i]} = f_{[i]}(1 + \epsilon_{[i]})$  and  $\underline{f}_{[j]} = f_{[j]}(1 + \epsilon_{[j]})$  approximate the Frobenius norms of some vectors of length  $i$  and  $j$ , respectively, and let*

$$\underline{f}_{[i,j]} = \text{hypot}(\underline{f}_{[i]}, \underline{f}_{[j]})$$

*be an approximation of the Frobenius norm of the concatenation (of length  $i+j$ ) of those two vectors. Then, barring any overflow and inexact underflow,*

$$\underline{f}_{[i,j]} = f_{[i,j]}(1 + \epsilon_{[i,j]}), \quad 1 + \epsilon_{[i,j]} = \sqrt{1 + \epsilon_{[i]}(2 + \epsilon_{[i]}) \frac{f_{[i]}^2}{f_{[i,j]}^2} (1 + \epsilon_{[j]}) (1 + \epsilon'_{[i,j]})}, \quad (8)$$

where  $1 + \epsilon_{[l]} = \min\{1 + \epsilon_{[i]}, 1 + \epsilon_{[j]}\}$ ,  $1 + \epsilon_{[k]} = \max\{1 + \epsilon_{[i]}, 1 + \epsilon_{[j]}\}$ , and  $1 + \epsilon_{[\ell]} = (1 + \epsilon_{[l]})/(1 + \epsilon_{[k]})$ , i.e.,  $l = i$  and  $k = j$  or  $l = j$  and  $k = i$ , while  $|\epsilon'_{[i,j]}| \leq \varepsilon$  if `hypot` is `cr_hypot`.

*Proof.* Expanding  $\underline{f}_{[i]}^2 + \underline{f}_{[j]}^2$  gives

$$\underline{f}_{[i]}^2 + \underline{f}_{[j]}^2 = \underline{f}_{[i]}^2(1 + \epsilon_{[i]})^2 + \underline{f}_{[j]}^2(1 + \epsilon_{[j]})^2 = (\underline{f}_{[l]}^2(1 + \epsilon_{[\ell]})^2 + \underline{f}_{[k]}^2)(1 + \epsilon_{[k]})^2. \quad (9)$$

Similarly to (5), expressing the first factor on the right hand side of (9) as

$$\underline{f}_{[l]}^2(1 + \epsilon_{[\ell]})^2 + \underline{f}_{[k]}^2 = (\underline{f}_{[l]}^2 + \underline{f}_{[k]}^2)(1 + z) \quad (10)$$

leads to

$$z = \epsilon_{[\ell]}(2 + \epsilon_{[\ell]}) \frac{\underline{f}_{[l]}^2}{\underline{f}_{[l]}^2 + \underline{f}_{[k]}^2} = \epsilon_{[\ell]}(2 + \epsilon_{[\ell]}) \frac{\underline{f}_{[l]}^2}{\underline{f}_{[i,j]}^2},$$

and therefore, by substituting (10) into (9),

$$\underline{f}_{[i,j]} = \sqrt{\underline{f}_{[i]}^2 + \underline{f}_{[j]}^2}(1 + \epsilon'_{[i,j]}) = \sqrt{\underline{f}_{[i]}^2 + \underline{f}_{[j]}^2} \sqrt{1 + z}(1 + \epsilon_{[k]})(1 + \epsilon'_{[i,j]}),$$

what is equivalent to (8).  $\square$

Compare this to computing the sum of squares, as in `xNRM2` from BLAS (see, e.g., <https://github.com/Reference-LAPACK/lapack/blob/master/BLAS/SRC/dnrm2.f90>).

**Theorem 3.** Let  $\mathbf{x} = [x_1 \ \cdots \ x_n]^T$  be a vector of finite floating-point values, and  $\|\mathbf{x}\|_F$  its Frobenius norm. If its approximation  $\|\mathbf{x}\|_F$  is computed as  $\|\mathbf{x}\|_F = \text{sqrt}(\underline{g}_n)$ , where  $\underline{g}_0 = g_0 = 0$  and

$$1 \leq i \leq n \implies \underline{g}_i = \text{fma}(x_i, x_i, \underline{g}_{i-1}),$$

then, barring any overflow and inexact underflow, when  $x_i \neq 0$  it holds

$$\underline{g}_i = g_i(1 + \epsilon''_i), \quad 1 + \epsilon''_i = \left(1 + \epsilon''_{i-1} \frac{g_{i-1}}{g_i}\right)(1 + \epsilon'''_i), \quad (11)$$

where  $\epsilon'''_i$ ,  $|\epsilon'''_i| \leq \varepsilon$ , is the rounding error of the `fma`. Also, with  $|\epsilon_{\sqrt{\cdot}}| \leq \varepsilon$ ,

$$\|\mathbf{x}\|_F = \text{sqrt}(\underline{g}_n) = \|\mathbf{x}\|_F \sqrt{1 + \epsilon''_n}(1 + \epsilon_{\sqrt{\cdot}}).$$

*Proof.* From

$$x_i^2 + \underline{g}_{i-1} = x_i^2 + g_{i-1}(1 + \epsilon''_{i-1}) = (x_i^2 + g_{i-1})(1 + w)$$

it follows

$$w = \epsilon''_{i-1} \frac{g_{i-1}}{g_{i-1} + x_i^2} = \epsilon''_{i-1} \frac{g_{i-1}}{g_i},$$

what had to be proven.  $\square$

## References

- [1] Laurent Fousse, Guillaume Hanrot, Vincent Lefèvre, Patrick Pélissier, and Paul Zimmermann. MPFR: A multiple-precision binary floating-point library with correct rounding. *ACM Trans. Math. Softw.*, 33(2):13, 2007.
- [2] OpenMP Architecture Review Board. OpenMP API 6.0 Specification. online: <https://www.openmp.org/wp-content/uploads/OpenMP-API-Specification-6-0.pdf>, Nov 2024.