

Title: Highly accurate EMRI self-force calculations

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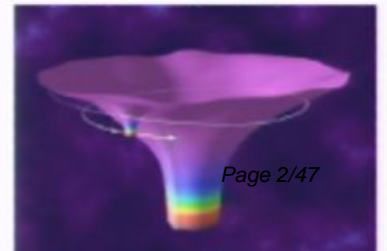
URL: <http://pirsa.org/10060041>

Abstract: TBA

**Highly accurate and efficient
self-force computation
using time-domain methods:
Error estimates, validation, and
optimization**

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Why High Accuracy: LISA (1)

Matched filtering of the entire LISA data stream is impractically expensive, but once sources have been detected by hierarchical and/or time-frequency searches, matched filtering will be used to estimate parameters for source modelling, cosmology, GR tests, . . .

Why High Accuracy: LISA (2)

High LISA S/N after matched filtering

⇒ Need highly accurate calculation of GW phase over the full LISA timespan ($\sim 10^5$ EMRI orbits) to prevent template-phase errors from dominating the overall parameter-estimation error budget

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The **strongest** LISA EMRI sources will have quite large S/N after matched filtering:

a	$\mathcal{R}_{\text{MW}}^{\text{BH}}$ (Gyr^{-1})	LISA Performance	
		5yr,2chan	2yr,1chan
0	4	140	16*
	40	460	180
	400	1300	490
0.5	4	160	20*
	40	560	210
	400	1500	590
0.9	4	240	46*
	40	780	330
	400	2100	860

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- ⇒ Need highly accurate calculation of GW phase over the full LISA timespan ($\sim 10^5$ EMRI orbits) to prevent template-phase errors from dominating the overall parameter-estimation error budget
- ⇒ Need highly accurate calculation of self-force: Huerta & Gair, PRD 79 (2009), 084021 find that a 10^{-6} fractional change in the EMRI self-force changes the cumulative GW phase by ~ 2 radians. Hence:

	Self-Force Relative Error Tolerance		
	$C = 1$	$C = 30$	$C = 1000$
$\rho_{\max} = 30$	2×10^{-8}	5×10^{-7}	2×10^{-5}
$\rho_{\max} = 300$	2×10^{-9}	5×10^{-8}	2×10^{-6}
$\rho_{\max} = 2000$	3×10^{-10}	8×10^{-9}	3×10^{-7}

[C measures the degeneracy (ill-conditioning) of the parameter estimation.]

Why High Accuracy: Orbit Correction

We can use the orbit-correction scheme of [Gralla & Wald, CQG 25 \(2008\), 205009](#) to explore $\mathcal{O}(\mu^2)$ effects, where $\mu = m/M \sim 10^{-5}$ is the EMRI mass ratio.

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To reliably distinguish true $\mathcal{O}(\mu^2)$ effects due to the orbit correction from numerical errors in the $\mathcal{O}(\mu)$ self force, the self force needs to be calculated with a relative error $\ll \mu$.

Other uses of highly accurate self-force calculations:

- calibrate/constraint post-Newtonian expansions (e.g., [Blanchet *et al.*, arXiv:0910.0207](#))
- test correctness of self-force calculations & theory by comparisons between calculations using different methods (e.g., [Sago, Barack, & Detweiler, PRD 78 \(2008\), 124024](#))

Why High Efficiency: Orbit Correction

The orbit-correction scheme of [Gralla & Wald, CQG 25 \(2008\), 205009](#) requires time-integrating a set of coupled ODEs for the orbital-parameter evolution on radiation-reaction and longer timescales, with the ODEs' right-hand-side functions being given by a self-force computation.

Even the most efficient ODE-integration schemes will require evaluating the right-hand-side functions (i.e., computing the self-force for some specified intermediate orbit) **hundreds of times** in the course of a single low-accuracy orbit-correction calculation, and **many more** for higher accuracies.

⇒ Self-force calculation should be **as efficient as possible!**

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⇒ Self-force calculation should be **as efficient as possible!**

[For comparison, [Barack & Sago, PRD 81 \(2010\), 084021](#) report that their gravitational self-force calculation takes 4–8 cpu days at a 10^{-4} relative accuracy level.]

Barack-Ori mode-sum regularization [**conceptual**]

Lorenz gauge \Rightarrow metric perturbation from point particle is relatively simple (locally isotropic).

Work in **time domain** \Rightarrow allows arbitrary particle orbits.

Decompose metric perturbation into spherical harmonics (ℓ, m) , compute each (ℓ, m) mode via time-domain **numerical integration** of the complex wave equation

$$\square \phi_{\ell m} + V_{\ell}(r) \phi_{\ell m} = S_{\ell m}(t) \delta(r(t) - r_{\text{particle}}(t))$$

Compute regularized self-force modes for each ℓ :

$$F_{\ell, \text{reg}}^{(\pm)} \sim \left(\sum_m (\nabla \phi_{\ell m}) \Big|_{r=r_p^{\pm}} \right) \mp (\ell + \frac{1}{2}) \underset{\substack{\uparrow \\ \text{(known coefficients)}}}{A} - \underset{\substack{\uparrow \\ \text{(known coefficients)}}}{B}$$

Compute self-force:

$$F_{\text{self}} = \sum_{\ell=0}^{\infty} F_{\ell}^{(\pm)}$$

Tail Force

The tail force $F_{\text{self,tail}}$ can be estimated via a “tail fit” of some subset of the $F_{\ell,\text{reg}}$ to the large- ℓ series

$$F_{\ell,\text{reg}} = \sum_{\substack{p \text{ even} \\ p \geq 2}} c_p f_p(\ell)$$

where the basis functions $f_p(\ell) = \mathcal{O}(\ell^{-p})$ are

$$f_2(\ell) = 1 / \left(\left(\ell - \frac{1}{2} \right) \left(\ell + \frac{3}{2} \right) \right)$$

$$f_4(\ell) = 1 / \left(\left(\ell - \frac{3}{2} \right) \left(\ell - \frac{1}{2} \right) \left(\ell + \frac{3}{2} \right) \left(\ell + \frac{5}{2} \right) \right)$$

$$f_6(\ell) = 1 / \left(\left(\ell - \frac{5}{2} \right) \left(\ell - \frac{3}{2} \right) \left(\ell - \frac{1}{2} \right) \left(\ell + \frac{3}{2} \right) \left(\ell + \frac{5}{2} \right) \left(\ell + \frac{7}{2} \right) \right)$$

...

Then $F_{\text{self,tail}} = \sum_p c_p \Gamma_p$ where $\Gamma_p = \sum_{\ell=K+1}^{\infty} f_p(\ell)$ can be computed analytically.

Adaptive Mesh Refinement (AMR)

I use **characteristic AMR** with (\geq) **4th order finite differencing** to numerically integrate the complex wave equation.

- Typically **speeds up computation by a factor of $\sim 30-40$** over an equivalent-resolution (time-domain) unigrid computation.

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- CPU time now scales \sim length of particle world-line (integration is cheap outside neighborhood of worldline).
 - \Rightarrow CPU time scales **linearly** with domain size for 1+1-D code
 - \Rightarrow relatively cheap to use very large domains for high accuracy (including waiting for low- ℓ initial-data transients to decay).

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 - \Rightarrow CPU time scales **linearly** with domain size for 1+1-D code
 - \Rightarrow relatively cheap to use very large domains for high accuracy (including waiting for low- ℓ initial-data transients to decay).
- Turn on AMR **gradually** \Rightarrow don't waste high resolution on resolving the initial-data transient.

See [Thornburg, arXiv:0909.0036](#) for details & **source code (C++)**.

Floating-Point Arithmetic

In some cases, my highest-accuracy results are limited in accuracy by floating-point rounding errors in solving the complex wave equation.

To investigate these effects, I compare results computed using standard IEEE “double” (64-bit) floating-point arithmetic versus IEEE “double-extended” (80-bit) floating-point arithmetic. Double-extended has a factor of $2^{11} = 2048$ smaller rounding errors, but is a factor of ~ 2 slower.

Test Parameters

Scalar-field particle in circular geodesic orbit in Schwarzschild spacetime at $r_{\text{areal}} = 10M$ (orbital period $\approx 199M$).

Double precision:

AMR error tolerance 10^{-14} , 10^{-15} , 10^{-16} ,
FMR may be none or $\times 2$, $\times 3$, $\times 4$, $\times 6$, or $\times 8$

Double-extended precision:

AMR error tolerance 10^{-14} down to 10^{-19} ,
FMR may be none or $\times 2$

Complex wave equation solved for $0 \leq \ell \leq 30$, $\ell = 35$, and $\ell = 40$.

Numerical force sums modes for $0 \leq \ell \leq K=30$.

Tail fit coefficients $\{c_4, c_6\}$ fitted to modes $20 \leq \ell \leq 30$.

[For a circular particle orbit c_2 is given analytically by

Detweiler, Messaritaki, & Whiting, PRD **67** (2003), 104016.]

Generic-orbit case ($\{c_0, c_1, c_2\}$ fitted) omitted here for brevity

Error Estimates for Individual Regularized Modes

Difference between $F_{l,\text{reg}}^{(+)}$ and $F_{l,\text{reg}}^{(-)}$ \Rightarrow “internal” error estimate:

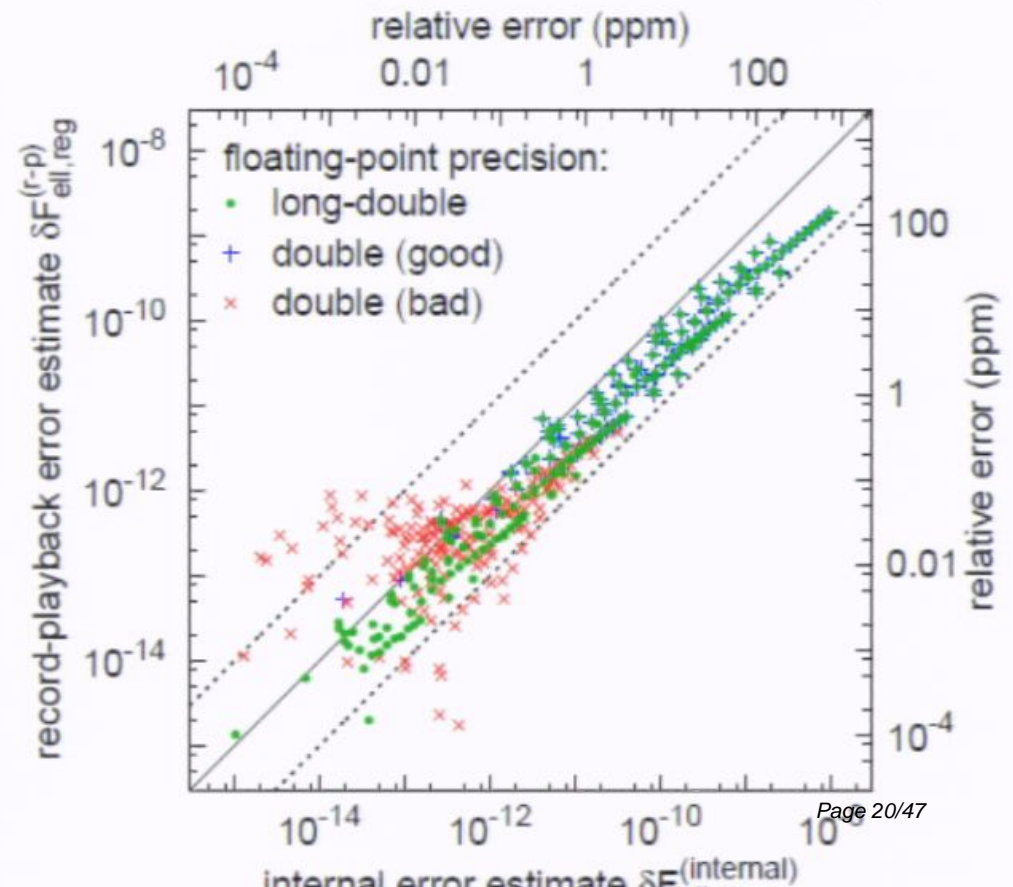
$$F_{l,\text{reg}} \pm \delta F_{l,\text{reg}}^{(\text{internal})} \equiv \frac{1}{2} \left(F_{l,\text{reg}}^{(+)} + F_{l,\text{reg}}^{(-)} \right) \pm \frac{1}{2} \left| F_{l,\text{reg}}^{(+)} - F_{l,\text{reg}}^{(-)} \right|$$

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Validate by comparing with convergence-test (“record-playback”) error estimates:



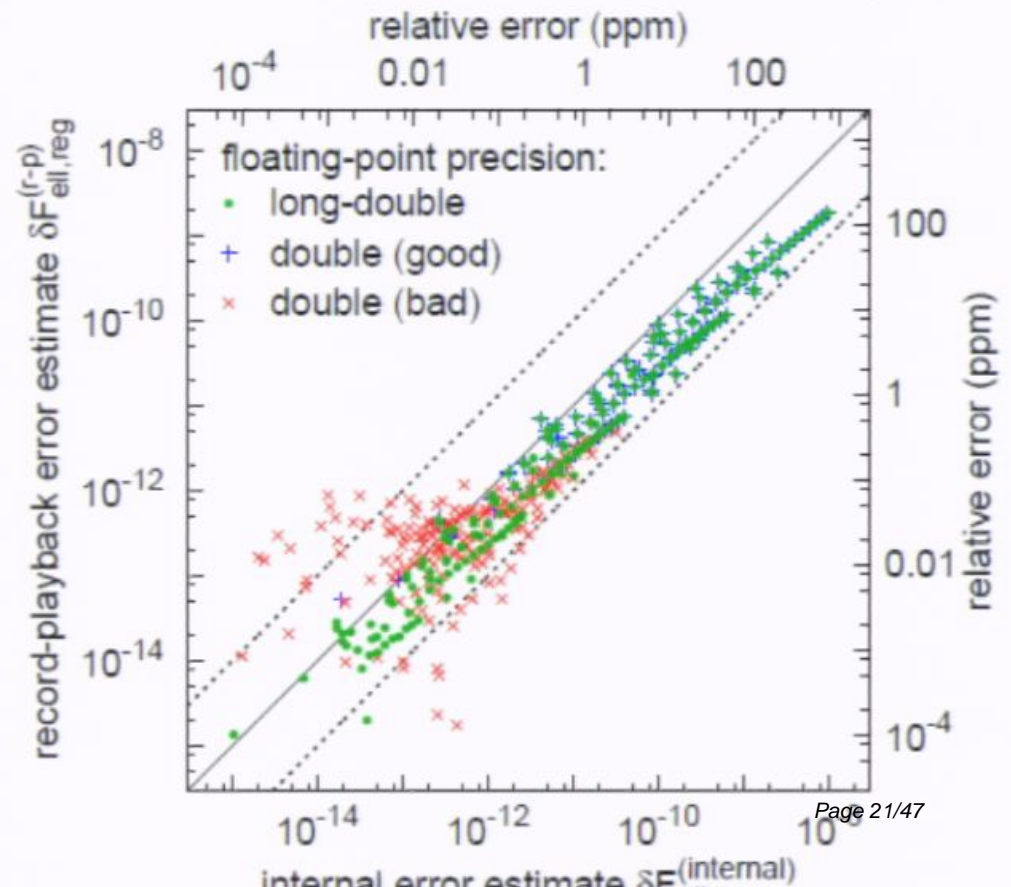
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Using **double-extended** floating point arithmetic, the internal error estimate is reliable & somewhat conservative, (pessimistic), overestimating the convergence-test errors by $\sim 1-10\times$.



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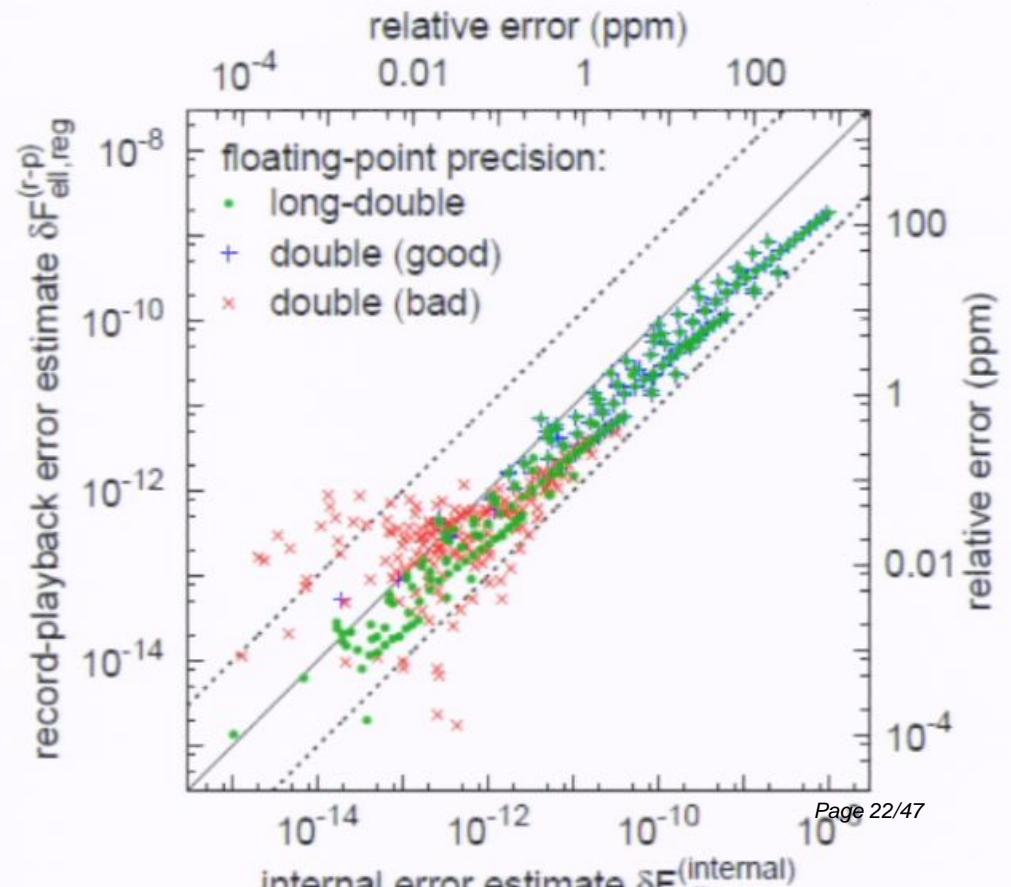
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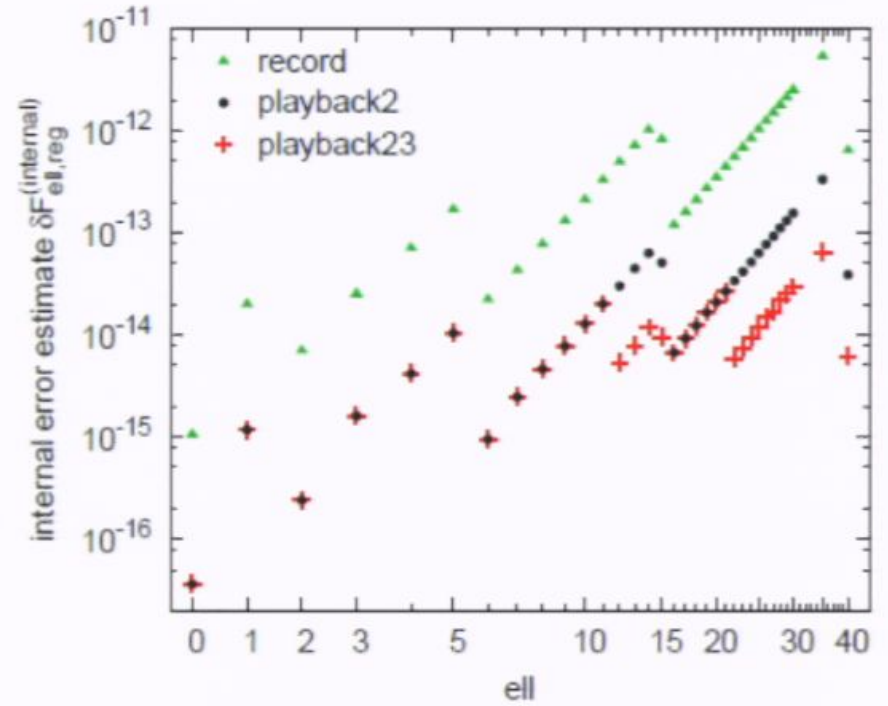
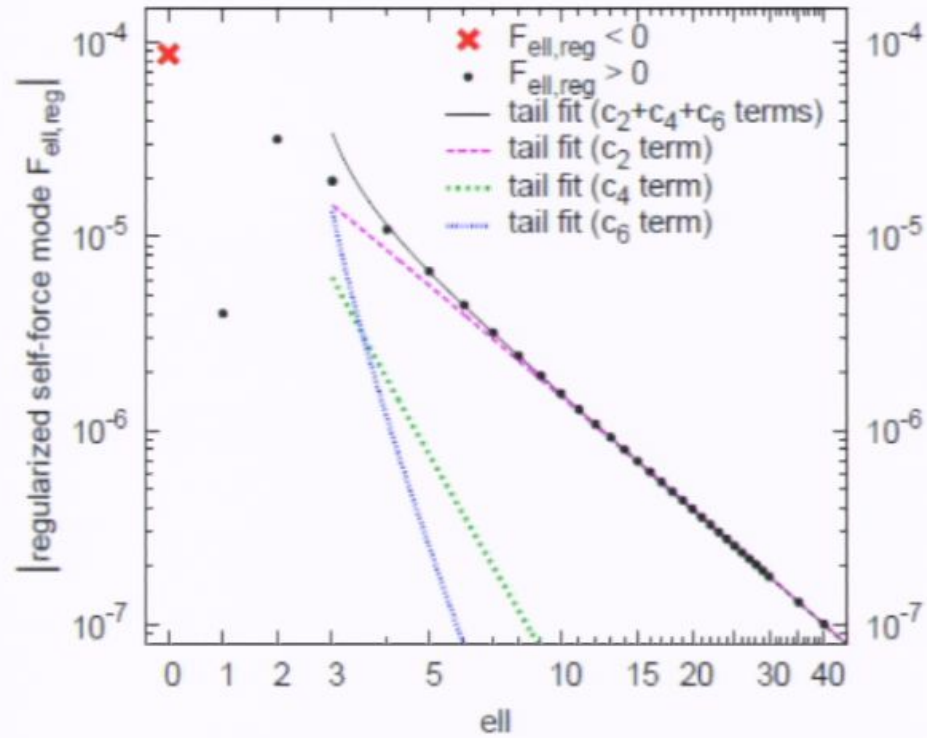
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Using standard “**double**” arithmetic, the internal error estimates is ok for $\delta F_{l,\text{reg}}^{(\text{internal})} \gtrsim 0.1$ ppm, but **unreliable** for $\delta F_{l,\text{reg}}^{(\text{internal})} \lesssim 0.1$ ppm due to floating-point rounding errors in solving the complex wave equation.



Overview of Modes and Error Estimates



Error Estimates for the Numerical Force (1)

We don't *a priori* know whether or to what extent the actual errors in the individual modes $F_{\ell, \text{reg}}$ for different ℓ might be correlated, so it's not obvious how to sum these to obtain an internal error estimate for the numerical force:

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Worst case: Actual numerical errors in different modes are perfectly correlated \Rightarrow should add individual modes' internal error estimates **arithmetically** [Barack & Sago, PRD 81 (2010), 084021 use this method].

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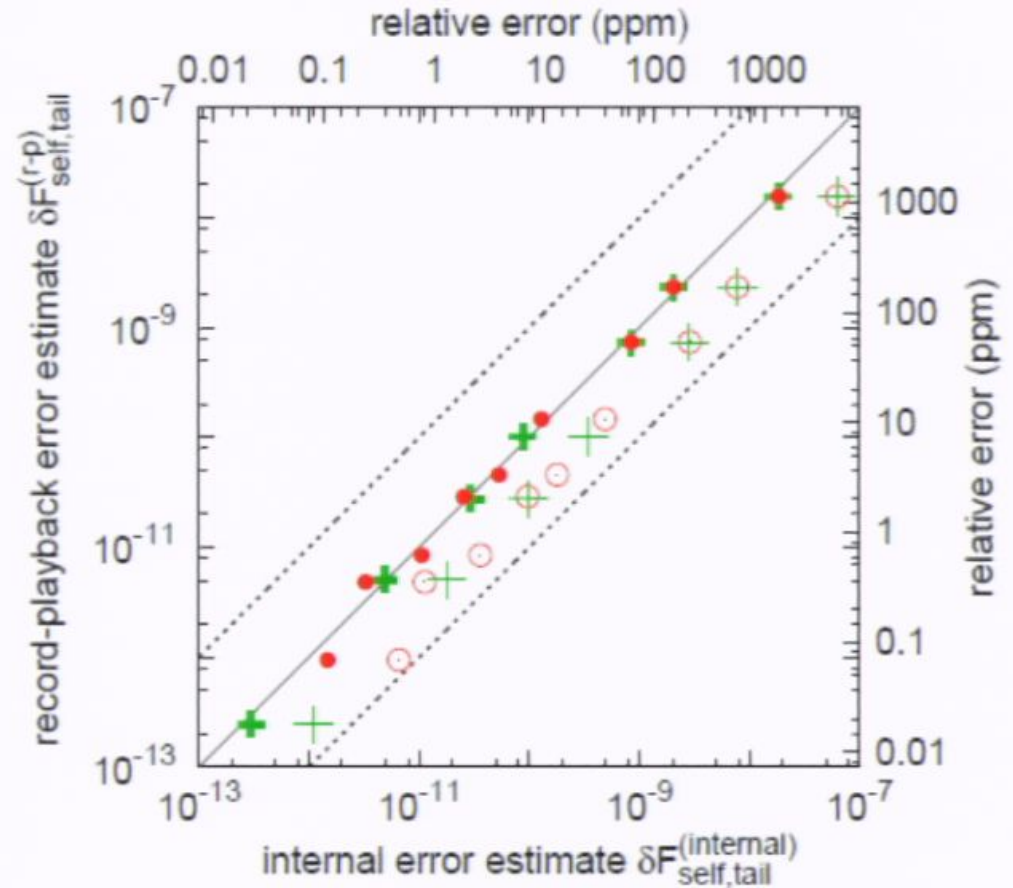
Best case: Actual numerical errors in different modes are (close enough to) statistically independent \Rightarrow should add individual modes' internal error estimates in **quadrature**.

Error Estimates for the Numerical Force (2)

Compute numerical-force internal error estimates $\delta F_{\text{self,num}}^{(\text{internal})}$ both ways (arithmetic & quadrature sums) and compare with convergence-test error estimates to see which works better:

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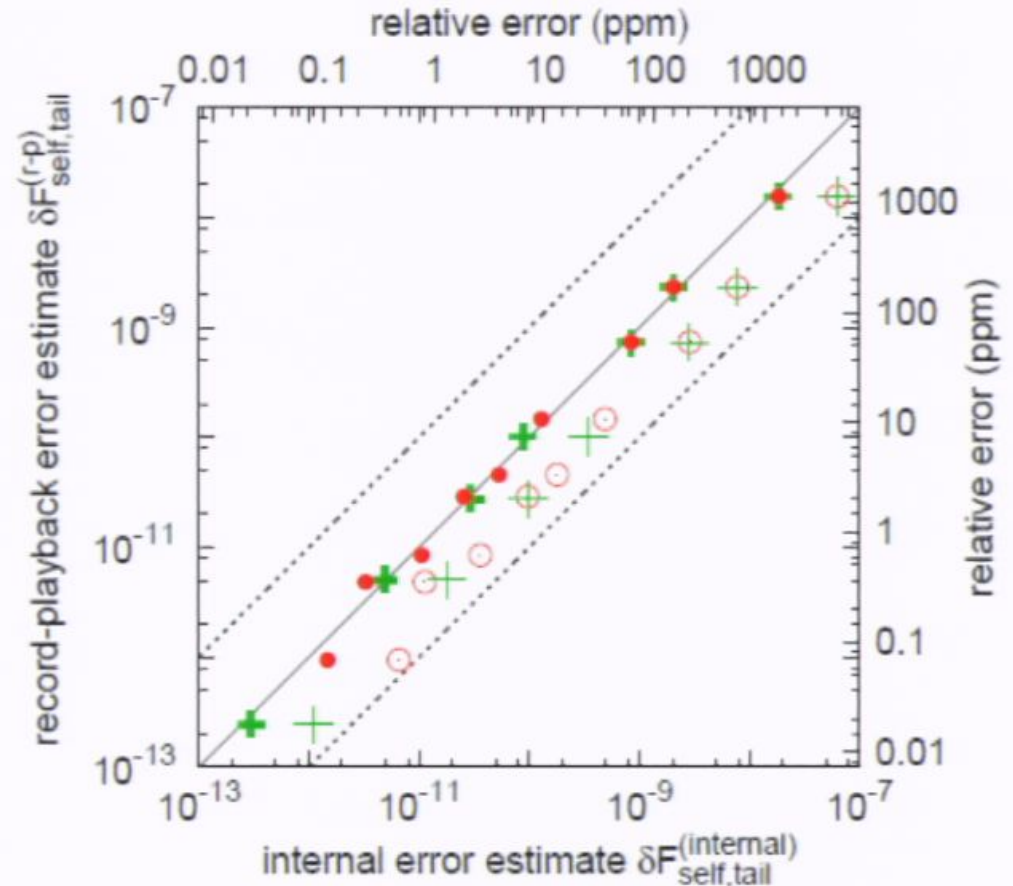
internal error estimate: ● double + long-double
 (33a) quadrature sum of modes ● +

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Compute numerical-force internal error estimates $\delta F_{\text{self,num}}^{(\text{internal})}$ both ways (arithmetic & quadrature sums) and compare with convergence-test error estimates to see which works better:

The arithmetic sum systematically overestimates the convergence-test error by $\sim 3\times$, while the quadrature sum gives an excellent estimate of the convergence-test error.

This is true for both double and double-extended arithmetic, even for errors so small that the individual modes' internal error estimates are unreliable! (This is apparently due to the averaging inherent in summing



internal error estimate: ● double + long-double
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II-Conditioning of the Tail Fit

The tail-fit basis functions $\{f_p\}$ are nearly linearly dependent, so **the tail fit is numerically ill-conditioned**. That is, there are linear combinations $\sum_p b_p f_p$ where the linear-combination coefficients $\{b_p\}$ have unit 2-norm, yet the linear combination $\sum_p b_p f_p$ is much smaller than the largest f_p .

Coefficients Being Fitted	Condition Number κ	
	Basis is $\{f_p\}$	Basis is $\{\bar{f}_p\}$
$\{c_4, c_6\}$	2.6×10^3	11
$\{c_4, c_6, c_8\}$	5.9×10^6	100
$\{c_2, c_4\}$	2.2×10^3	8.3
$\{c_2, c_4, c_6\}$	4.4×10^6	66
$\{c_2, c_4, c_6, c_8\}$	8.0×10^9	460

The ill-conditioning can be greatly reduced by renormalizing the **basis functions** to have similar magnitudes over the range of ℓ used

Error Estimates for the Tail Fit (1)

Again, we don't *a priori* know whether or to what extent the actual errors in the individual modes $F_{\ell,\text{reg}}$ for different ℓ might be correlated, so it's not obvious that we can apply the standard "statistics-101" formulae for error propagation in least-squares fitting:

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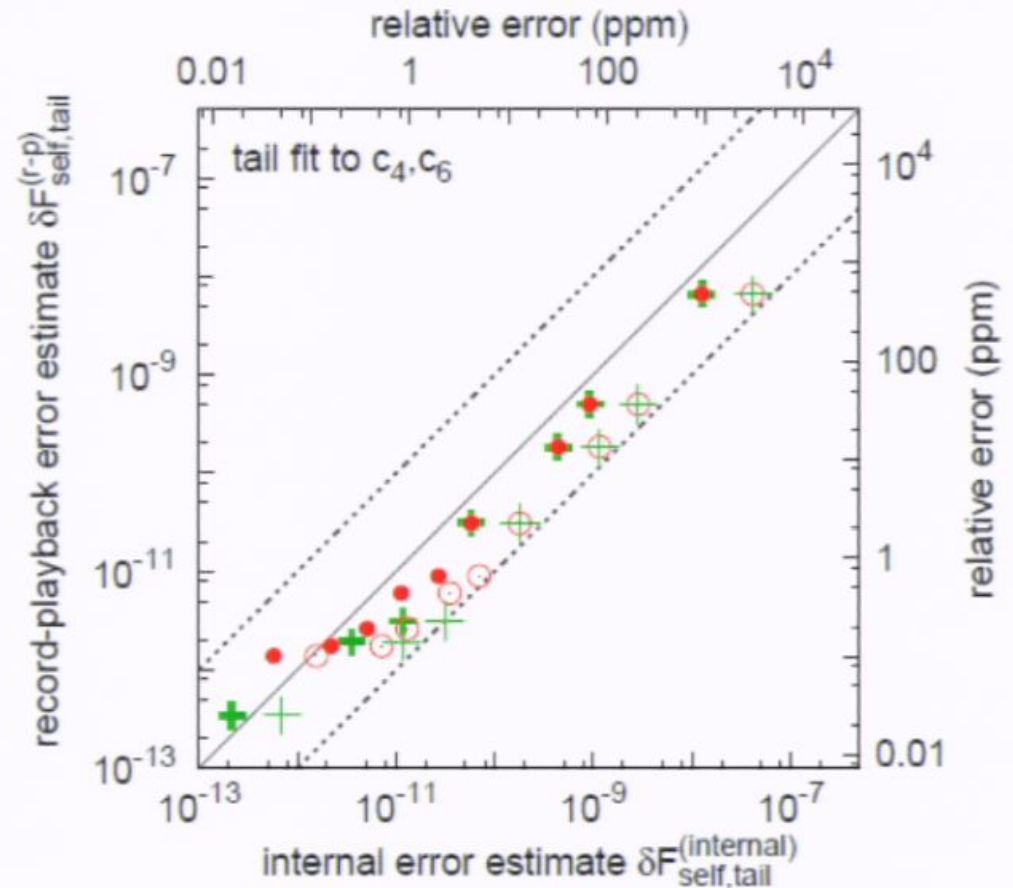
- Best case:** Actual numerical errors in different modes are (close enough to) statistically independent that we can use the standard **statistical error estimate**.
- Worst case:** Estimate self-force errors by trying a **trial fit for each possible combination of $\{-, 0, +\}$ signs** of the different modes' internal error estimates. (For Q modes in the fit, there are 3^Q combinations, and hence 3^Q trial fits are needed.)

Error Estimates for the Tail Fit (2)

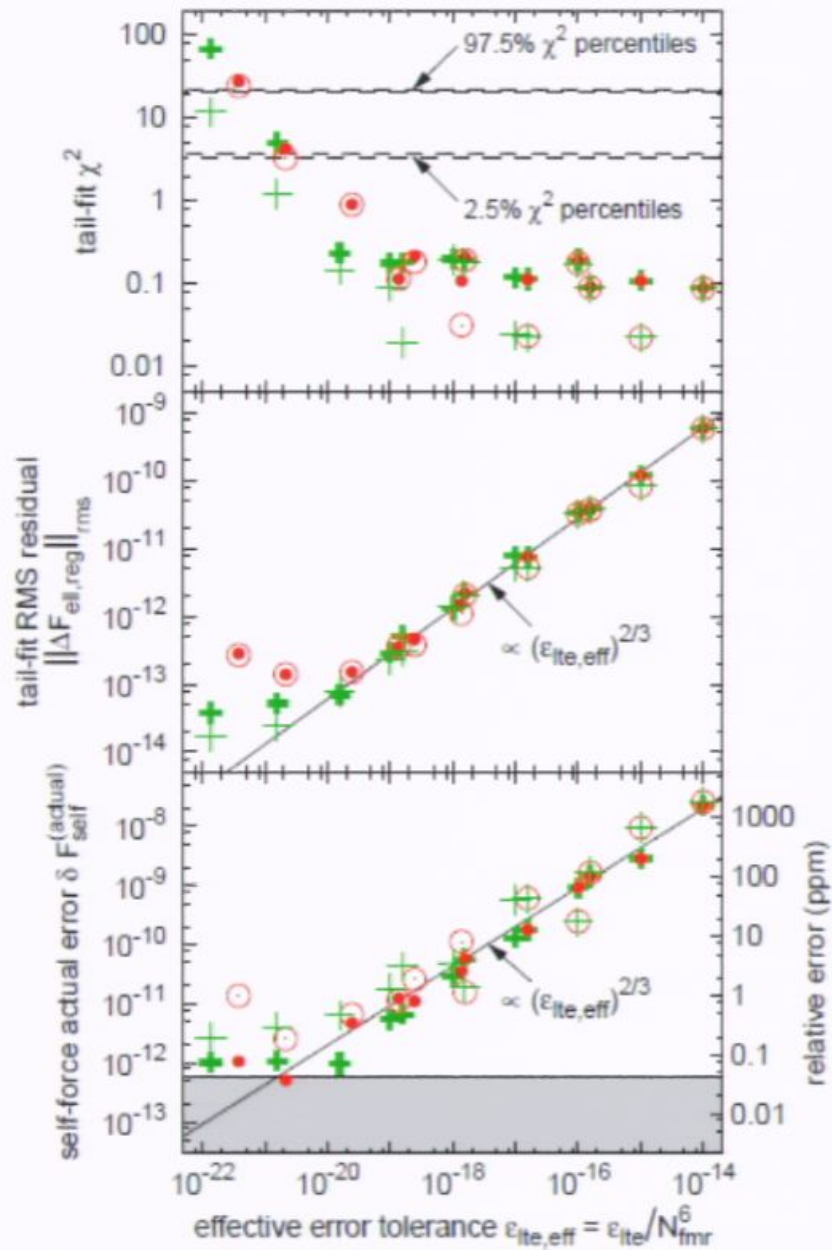
Compute tail-fit internal error estimates $\delta F_{\text{self,tail}}^{(\text{internal})}$ both ways (statistical and worst-case-of- 3^Q -combinations-of-error-signs) and compare with convergence-test error estimates to see which works better:

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Tail-Fit Results



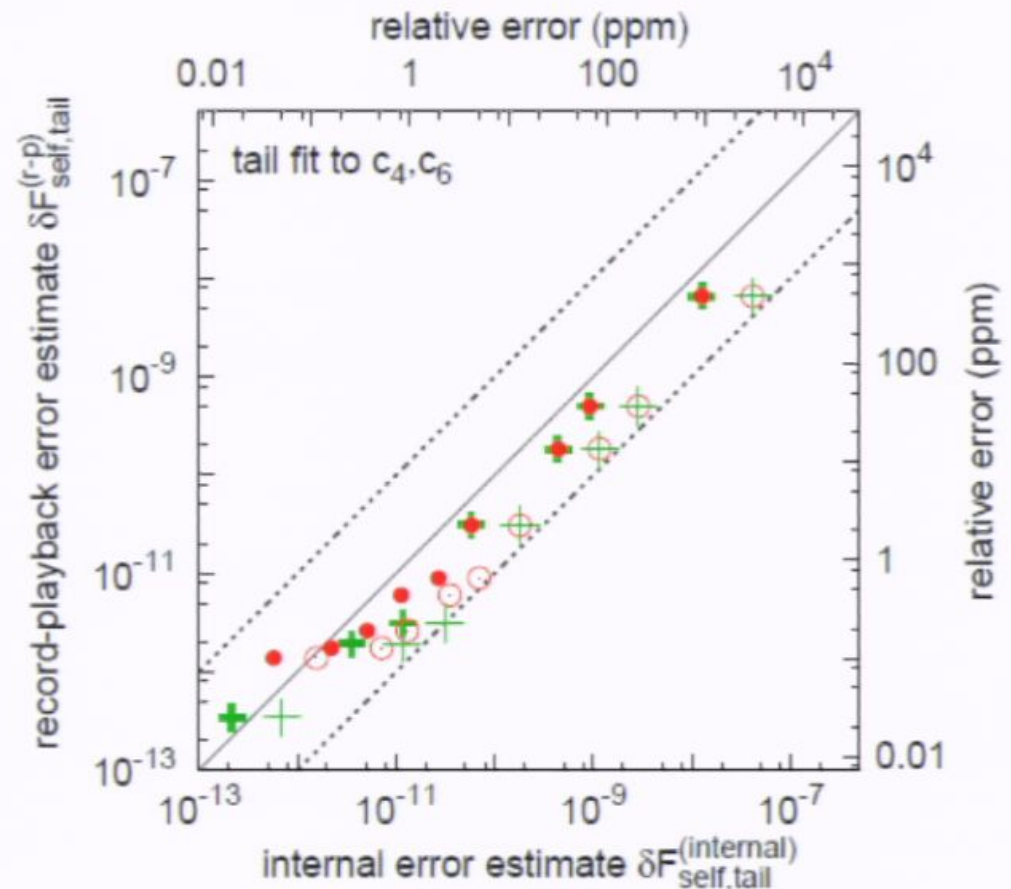
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The worst-case error estimate systematically overestimates the convergence-test error by $\sim 6\times$, while the statistical error estimate is a fairly good (conservative by $\sim 2\times$) estimate of the convergence-test error.



internal error estimate: double long-double
 (34a) statistical error in tail fit ● +

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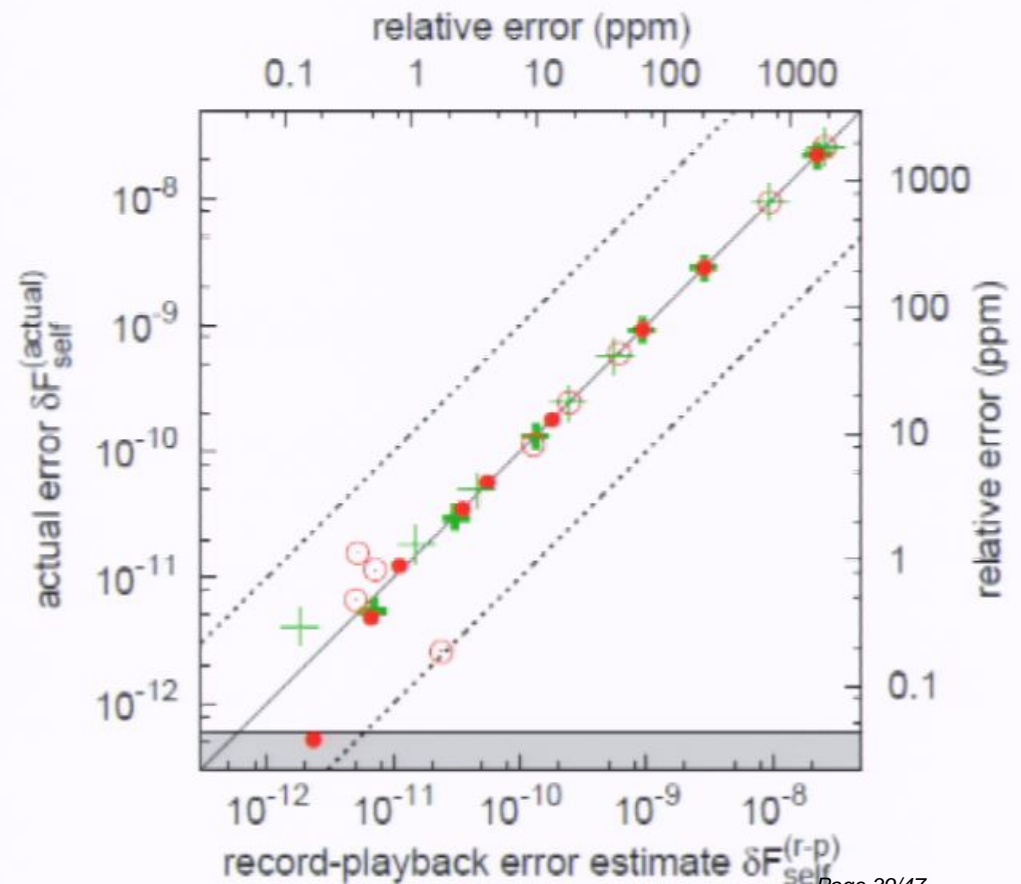
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Validate Overall Self-Force Error Estimates

Validate the convergence-test error estimates for the overall self-force [quadrature sum of numerical-force and tail-force error; details omitted for brevity] by comparing with highly-accurate frequency-domain results from Detweiler, Messaritaki, & Whiting, PRD 67 (2003), 104016:

Very good agreement except for high-accuracy results with double floating-point precision.

[Shaded region shows error $\leq 3 \times$ Detweiler, Messaritaki, & Whiting error estimate]



Error Estimates for the Self-Force

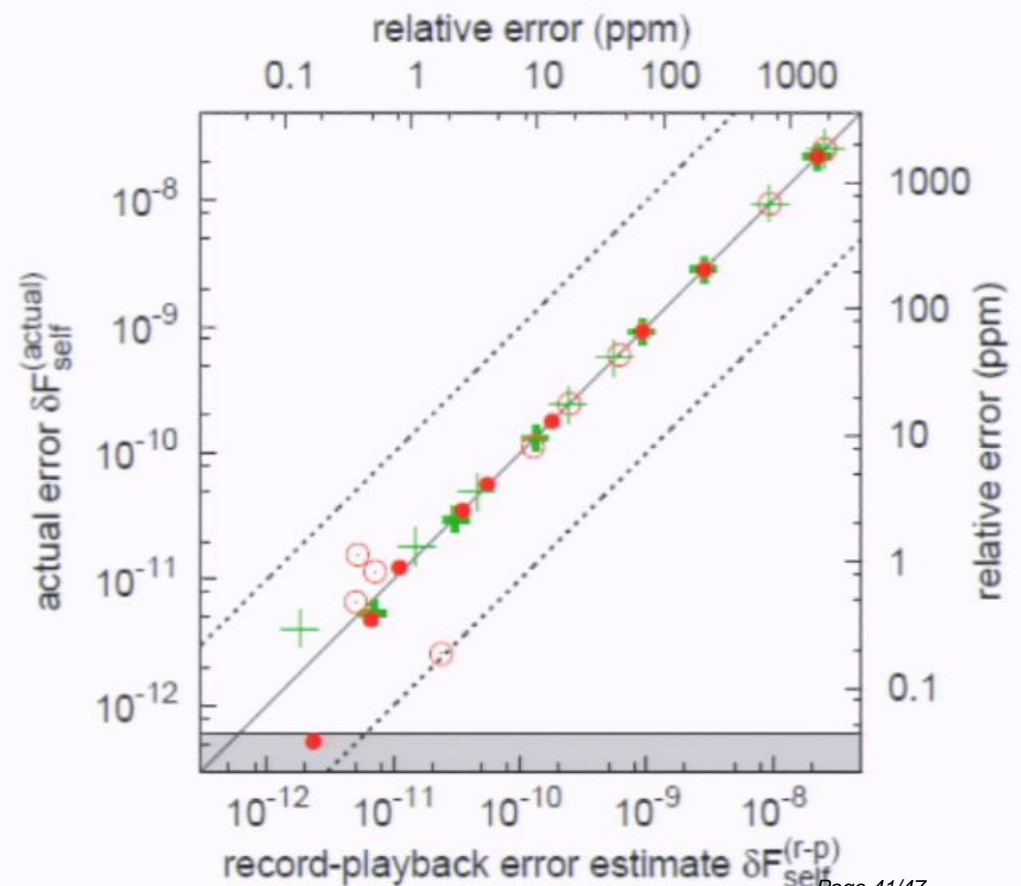
Compute this error estimate as **quadrature** sum of error estimates for numerical force & tail force [details of validation of quadrature vs. arithmetic sum omitted for brevity].

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tail fit coefficients:

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long-double

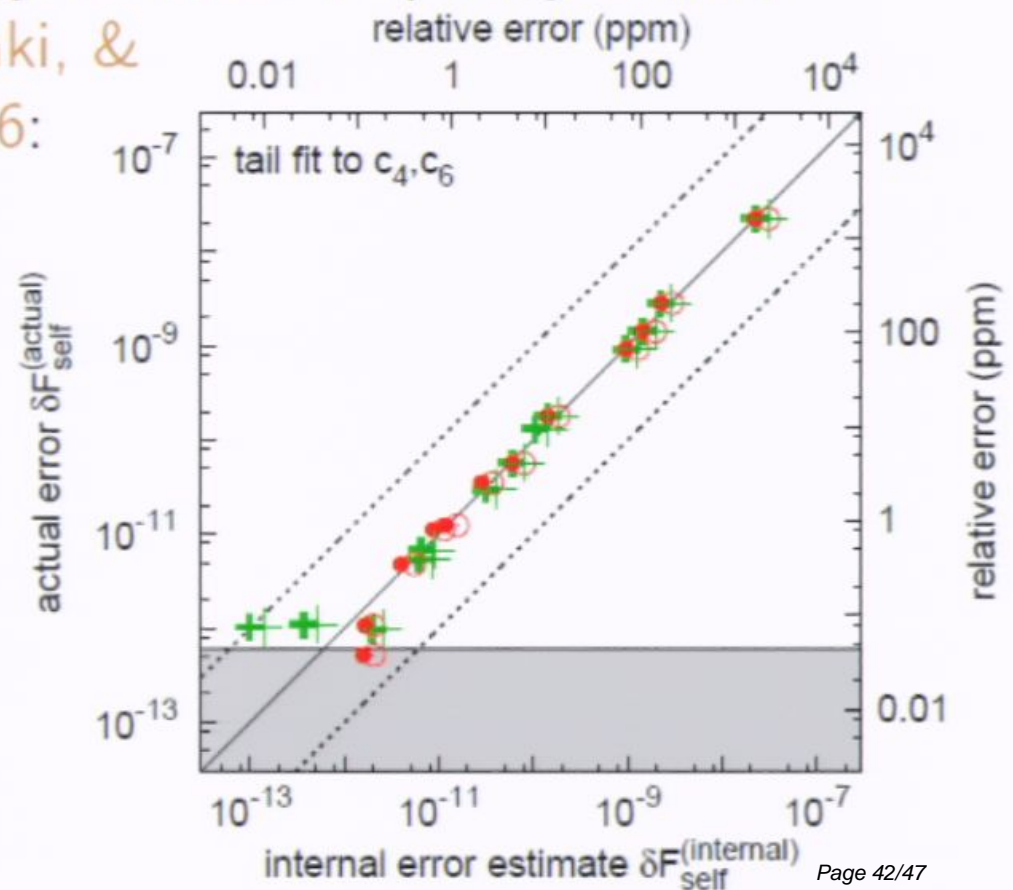
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Validate by comparing with highly-accurate frequency-domain results from **Detweiler, Messaritaki, & Whiting, PRD 67 (2003), 104016**:

Very good agreement except for errors $\lesssim 0.1$ ppm.

[Shaded region shows error $\leq 3 \times$ Detweiler, Messaritaki, & Whiting error estimate]



Sample Results

AMR error tolerance 10^{-19} (double-extended floating-point)

method	F_{self} (q^2/M)	$\delta F_{\text{self}}^{(\text{internal})}$ (ppm)	$\delta F_{\text{self}}^{(\text{actual})}$ (ppm)
record	$1.378\,448\,82 \times 10^{-5}$	0.44	0.39
playback2	$1.378\,448\,169\,8 \times 10^{-5}$	0.027	0.080
D., M., & W.	$1.378\,448\,28 \times 10^{-5}$	0.015	

CPU time is currently quite slow at more modest accuracy, because numerical parameters are still tuned for the very highest-accuracy test runs:

- ▶ Mostly same AMR error tolerance for each ℓ (more efficient would be less-accurate tolerance for modes with small $F_{\ell, \text{reg}}$)
- ▶ $\ell = 0$ integrations use a domain size of $30\,000M$ or $100\,000M$.
- ▶ Complex wave equation solved for $0 < \ell < 30$, $\ell = 35$, and

Conclusions

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- ▶ Internal error estimate from $|\text{outside} - \text{inside}|$.
- ▶ It's safe to treat different modes' internal error estimates as statistically independent for purposes of computing error estimates for the numerical force and tail force.
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- ▶ Tail fit is highly ill-conditioned when fitting many orders simultaneously; renormalizing basis functions mostly solves this problem.
- ▶ Floating-point roundoff limits my AMR code to ~ 0.1 ppm.
- ▶ My results agree beautifully with highly-accurate frequency-domain results of [Detweiler, Messaritaki, & Whiting, PRD 67 \(2003\), 104016](#).
⇒ confirms correctness of both codes & theories

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- ▶ It's safe to treat different modes' internal error estimates as statistically independent for purposes of computing error estimates for the numerical force and tail force.
- ▶ Tail fit is highly ill-conditioned when fitting many orders simultaneously; renormalizing basis functions mostly solves this problem.
- ▶ Floating-point roundoff limits my AMR code to ~ 0.1 ppm.
- ▶ My results agree beautifully with highly-accurate frequency-domain results of [Detweiler, Messaritaki, & Whiting, PRD **67** \(2003\), 104016](#).
⇒ confirms correctness of both codes & theories