

The need for accuracy and smoothness in numerical simulations

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This is not a talk about, say,

- ① computational speed
- ② parallelism
- ③ energy efficiency

This is a talk about

the relevance of the numbers returned by our simulations

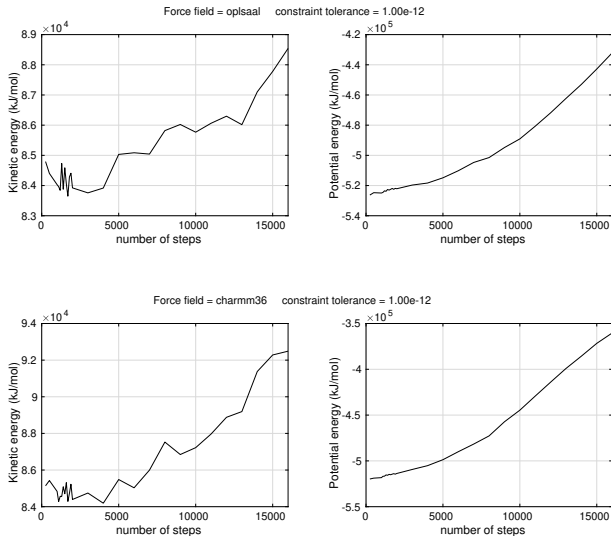
Warning: Disturbing pictures to follow

We did two sets of simulations of a molecule in water

- ① fixed real time
- ② two different force-fields: charmm36, oopsaal
- ③ identical initial conditions

and tracked the energy as a function of the number of steps.

Warning: Disturbing pictures below



The system of differential algebraic equations

$$\mathbf{q}'(t) = \mathbf{v}(t), \quad (1)$$

$$\mathbf{M}\mathbf{v}'(t) = \mathbf{f}(\mathbf{q}(t)) - \mathbf{G}(\mathbf{q}(t))^T \boldsymbol{\lambda}(t), \quad (2)$$

$$\mathbf{g}(\mathbf{q}(t)) = \mathbf{0}. \quad (3)$$

is solved using the SHAKE algorithm

$$\mathbf{v}_{n+1/2} = \mathbf{v}_{n-1/2} + h\mathbf{M}^{-1} \left(\mathbf{f}(\mathbf{q}_n) - \mathbf{G}(\mathbf{q}_n)^T \boldsymbol{\lambda}_n \right), \quad (4)$$

$$\mathbf{q}_{n+1} = \mathbf{q}_n + h\mathbf{v}_{n+1/2}, \quad (5)$$

$$\mathbf{g}(\mathbf{q}_{n+1}) = \mathbf{0}. \quad (6)$$

Key questions

Does it matter

- ① if the force-field

$$\mathbf{q} \rightarrow \mathbf{f}(\mathbf{q})$$

is smooth or not?

- ② if the nonlinear constraint equation

$$\mathbf{g}(\mathbf{q}_{n+1}(\boldsymbol{\lambda})) = \mathbf{0} \tag{7}$$

is solved accurately with respect to $\boldsymbol{\lambda}$ or not?

The primary point of this talk

If one of the following is true:

- ① the central functions are not smooth enough
- ② the central equations are not solved accurately enough

then we cannot:

- ① estimate the modelling error
- ② validate the model against the physical reality

The key terms of this talk

- ① P : physical quantity that can be measured
- ② T : approximation of P predicted by our model
- ③ A_h : approximation of T returned by our algorithm
- ④ \hat{A}_h : the computed value of A_h .

The three different error terms

- ① $P - T$ is the modelling error and

$$P - T \neq 0 \quad (8)$$

because our model is simpler than the real world.

- ② $T - A_h$ is the discretization error and

$$T - A_h \neq 0 \quad (9)$$

because we cannot solve most differential equations exactly.

- ③ $A_h - \hat{A}_h$ is the computational error and

$$A_h - \hat{A}_h \neq 0 \quad (10)$$

due to rounding errors and truncation errors.

Why should we care?

We validate our models by demonstrating that

$$P - T \approx 0 \quad (11)$$

is a good approximation, but

$$P - T \approx P - \hat{A}_h \quad (12)$$

is not necessarily a good approximation. We have

$$P - T = (P - \hat{A}_h) - (T - A_h) - (A_h - \hat{A}_h) \quad (13)$$

so we need to assert that

$$|T - A_h| \ll |P - \hat{A}_h| \quad (14)$$

$$|A_h - \hat{A}_h| \ll |P - \hat{A}_h| \quad (15)$$

Practical error estimation

It is frequently possible to simultaneously

- 1 assert that the computational error

$$A_h - \hat{A}_h$$

is irrelevant, and

- 2 estimate the discretization error

$$T_h - A_h$$

accurately

The key is the existence of an asymptotic error expansion

$$E_h = T - A_h = \alpha h^p + \beta h^q + O(h^r), \quad h \rightarrow 0_+ \quad (16)$$

where $0 < p < q < r$.

Elementary results

Define

$$R_h := \frac{A_h - A_{2h}}{2^p - 1}, \quad F_h := \frac{A_{2h} - A_{4h}}{A_h - A_{2h}} \quad (17)$$

If

$$E_h = T - A_h = \alpha h^p + \beta h^q + O(h^r), \quad h \rightarrow 0_+ \quad (18)$$

then

$$\frac{E_h - R_h}{h^q} \rightarrow \text{constant} \quad (19)$$

and

$$\frac{2^p - F_h}{h^{q-p}} \rightarrow \text{constant} \quad (20)$$

which implies

$$\log |2^p - F_h| \approx \log |\text{constant}| + (q - p) \log h \quad (21)$$

An example from elementary numerical analysis

Our target value is

$$T = \int_0^1 f(x) dx, \quad f(x) = \exp(x) \quad (22)$$

Our approximation is the composite trapezoidal rule

$$A_h = \frac{h}{2} \sum_{j=0}^{n-1} (f(x_j) + f(x_{j+1})), \quad x_j = jh, \quad nh = 1 \quad (23)$$

We compute A_h for

$$h = h_k = 2^{-k} \quad (24)$$

The evolution of F_h and the quality of R_h

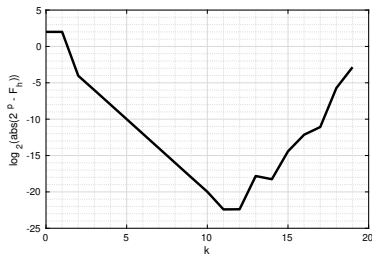


Figure: The evolution of F_h

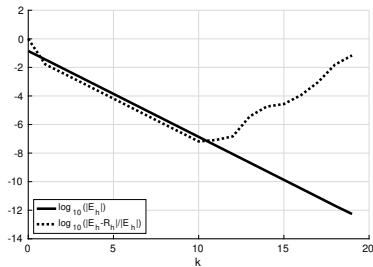


Figure: The accuracy of R_h

Strictly speaking we are not observing F_h and R_h

we are observing \hat{F}_h and \hat{R}_h

The discrepancy is controlled by the computational error

$$A_h - \hat{A}_h$$

A spectacular failure

GROMACS simulation of hen egg white lysozyme in water:

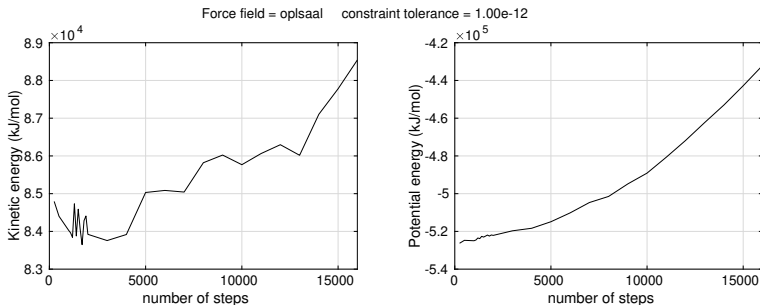
- 1 T : total energy of the simulation at the end
- 2 A_h : the approximation of T computed using SHAKE
- 3 \hat{A}_h : the value of A_h returned by the computer

Temporal matters:

- 1 The length of the simulations was 10^{-12} s (1 ps)
- 2 A common step size in MD is 10^{-15} s (1 fs)

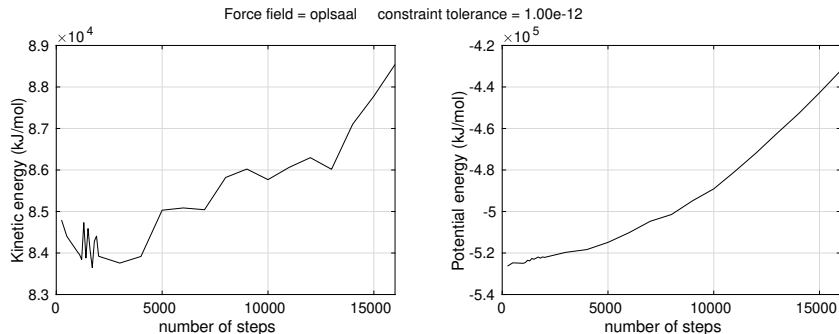
So, a mere 1000 steps! What could possibly go wrong?

A spectacular failure



The rapid growth of the energy for large value of n
can likely be cured using compensated summation.

A spectacular failure



- The wiggles near $n = 1000$ steps are a great concern.
- If there is an AEX, then 1 fs is not inside the asymp. range.
- We cannot assert that rounding errors are irrelevant
- We cannot estimate the discretisation error

What can go wrong?

- ① Proving the existence of an AEX is an exercise in
Taylor expansion

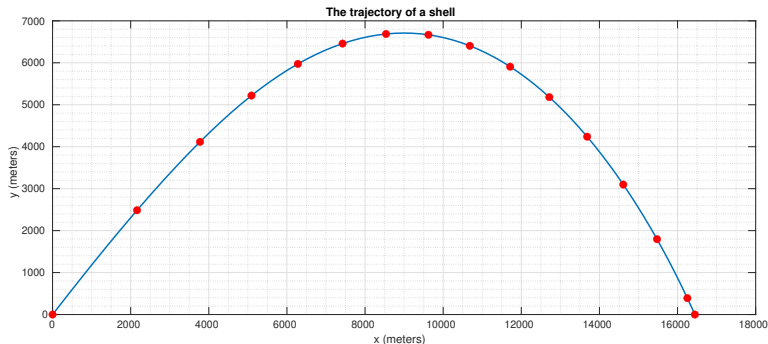
We cannot complete the proof unless
the functions are sufficiently smooth.

- ② We cannot expect \hat{A}_h to behave as predicted by the AEX
unless

$$\hat{A}_h \approx A_h$$

is a good approximation

External ballistics



The total force \mathbf{F} acting on the shell

$$\mathbf{F} = m\mathbf{g} - \frac{1}{2}\rho(y)AC_D(\nu, y)\|\mathbf{v} - \mathbf{w}\|_2(\mathbf{v} - \mathbf{w}) \quad (25)$$

is the combination of gravity and aerodynamic drag.

Computing the optimal range of the a howitzer

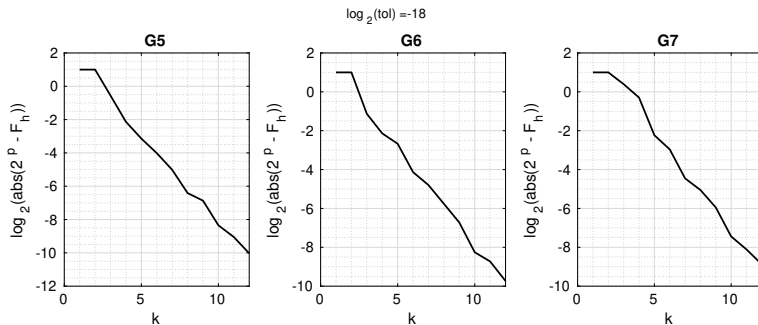


Figure: The evolution of Richardson's fraction for 3 different drag coefficients, Euler's explicit method and sufficiently accurate event location.

Computing the optimal range of the a howitzer

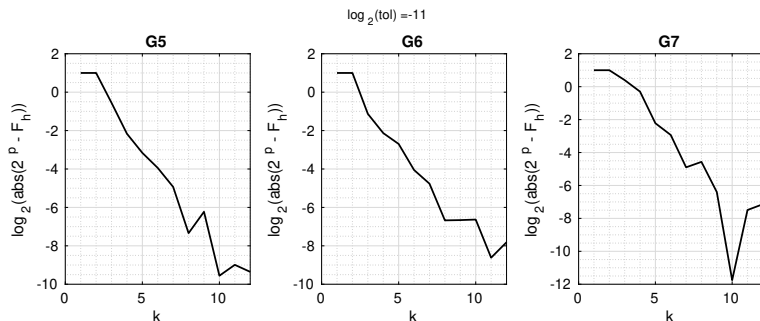


Figure: The evolution of Richardson's fraction for 3 different drag coefficients, Euler's explicit method and inaccurate event location.

Modelling ions

Why should this concern the PPAM community?

We use every trick in the book to

- ① reduce time-to-solution
- ② increase the parallel efficiency
- ③ reduce energy-to-solution

We should not forget to ask the question:

Did we just compromise our ability to validate our models against
the real world?