

falcON

a Cartesian FMM for the low-accuracy regime

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N-body simulations in astronomy



HCG87: a group of galaxies



ω Cen: a globular cluster

properties of stellar systems

- ▷ simple physics: Newtonian gravity
- ▷ very inhomogeneous
 - ⇒ large dynamic range
- ▷ dynamically young ($t_{\text{dyn}} \simeq \text{Myr} - \text{Gyr}$)
- ▷ well approximated as ensembles of point masses
 - ⇒ well described as Hamiltonian systems
 - (⇒ need symplectic time integration)

$$H = \sum_{i=1}^N \frac{m_i}{2} \left[v_i^2 - \sum_{j \neq i} \frac{G m_j}{|\mathbf{x}_i - \mathbf{x}_j|} \right], \quad \mathbf{v}_i = \dot{\mathbf{x}}_i = \frac{\mathbf{p}_i}{m_i}$$

with $N \simeq 10^{5-20}$

- ▷ equation of motion in continuum (mean-field) limit:

$$0 = \frac{df}{dt} = \frac{\partial f}{\partial t} + \frac{\partial f}{\partial \mathbf{x}} \cdot \mathbf{v} - \frac{\partial f}{\partial \mathbf{v}} \cdot \frac{\partial \Phi}{\partial \mathbf{x}}$$

collisionless Boltzmann equation (CBE)

- ▷ $f(\mathbf{x}, \mathbf{v}, t)$: **distribution function** (density in phase space)
- ▷ $\Phi(\mathbf{x})$: mean-field **gravitational potential**

both are related via the **Poisson equation**:

$$\nabla^2 \Phi(\mathbf{x}) = 4\pi G \int d^3v f(\mathbf{x}, \mathbf{v}, t)$$

two-body relaxation

How good is the continuum description?

- ▷ stellar encounters deflect trajectories
 - ⇒ stellar orbits get randomized
 - ⇒ Maxwellian velocity distribution

▷ **two-body relaxation time:**

$$t_{\text{relax}} \simeq 0.1 \frac{N}{\log N} t_{\text{dyn}}$$

1 collision-dominated stellar dynamics

- ▷ $t_{\text{relax}} \lesssim$ age of system
 - ⇒ continuum limit not applicable
 - ⇒ must simulate Hamiltonian directly:
 - ▷ force computation is $\mathcal{O}(N^2)$
 - ⇒ computational effort limits $N \lesssim 10^5$
 - ▷ close encounters are important
 - ⇒ time integration becomes tedious

2 collisionless stellar dynamics

- ▷ $t_{\text{relax}} \gg$ age of system
 - ⇒ continuum limit applicable
 - ⇒ solve **CBE & Poisson equation**

'collisionless' N-body simulations

How to solve the CBE?

$$0 = \frac{df}{dt} = \frac{\partial f}{\partial t} + \frac{\partial f}{\partial \mathbf{x}} \cdot \mathbf{v} - \frac{\partial f}{\partial \mathbf{v}} \cdot \frac{\partial \Phi}{\partial \mathbf{x}}$$

- ▷ f is 6D & very inhomogeneous
 - ⇒ (Eulerian) grid methods are useless
 - ⇒ Lagrangian method ('method of characteristics'):
- ▷ sample N trajectories $\{\mu_i, \mathbf{x}_i, \mathbf{v}_i\}$ from $f(\mathbf{x}, \mathbf{v}, t = 0)$
- ▷ solve equations of motion $\ddot{\mathbf{x}}_i = -\nabla \Phi(\mathbf{x}_i, t)$
- ▷ CBE: $\mu_i = \text{const}$ along trajectories
 - ⇒ $f(\mathbf{x}, \mathbf{v}, t)$ is represented by $\{\mu_i, \mathbf{x}_i(t), \mathbf{v}_i(t)\}$
 - ⇒ f is *unknown*
 - ⇒ **moments** of f can be *estimated*
 - ⇒ $N \ll N$ is *numerical parameter*
 - ⇒ artificial two-body relaxation

How to solve the Poisson equation?

$$\nabla^2 \Phi(\mathbf{x}) = 4\pi G \int d^3v f(\mathbf{x}, \mathbf{v}, t)$$

1 grid techniques (FFT, multigrid):

- ▷ fast: $\mathcal{O}(n_{\text{grid}} \log n_{\text{grid}})$
- ▷ periodic (\Rightarrow cosmology)
- ▷ problem: inhomogeneity (but: adaptive multigrid)

2 basic functions (using Y_{lm}):

- ▷ fast: $\mathcal{O}(N n_{\text{basis}})$
- ▷ problems: central singularity, spherical symmetry

3 Greens-function approach:

$$\Phi(\mathbf{x}, t) = -G \int d^3x' d^3v \frac{f(\mathbf{x}', \mathbf{v}, t)}{|\mathbf{x} - \mathbf{x}'|}$$

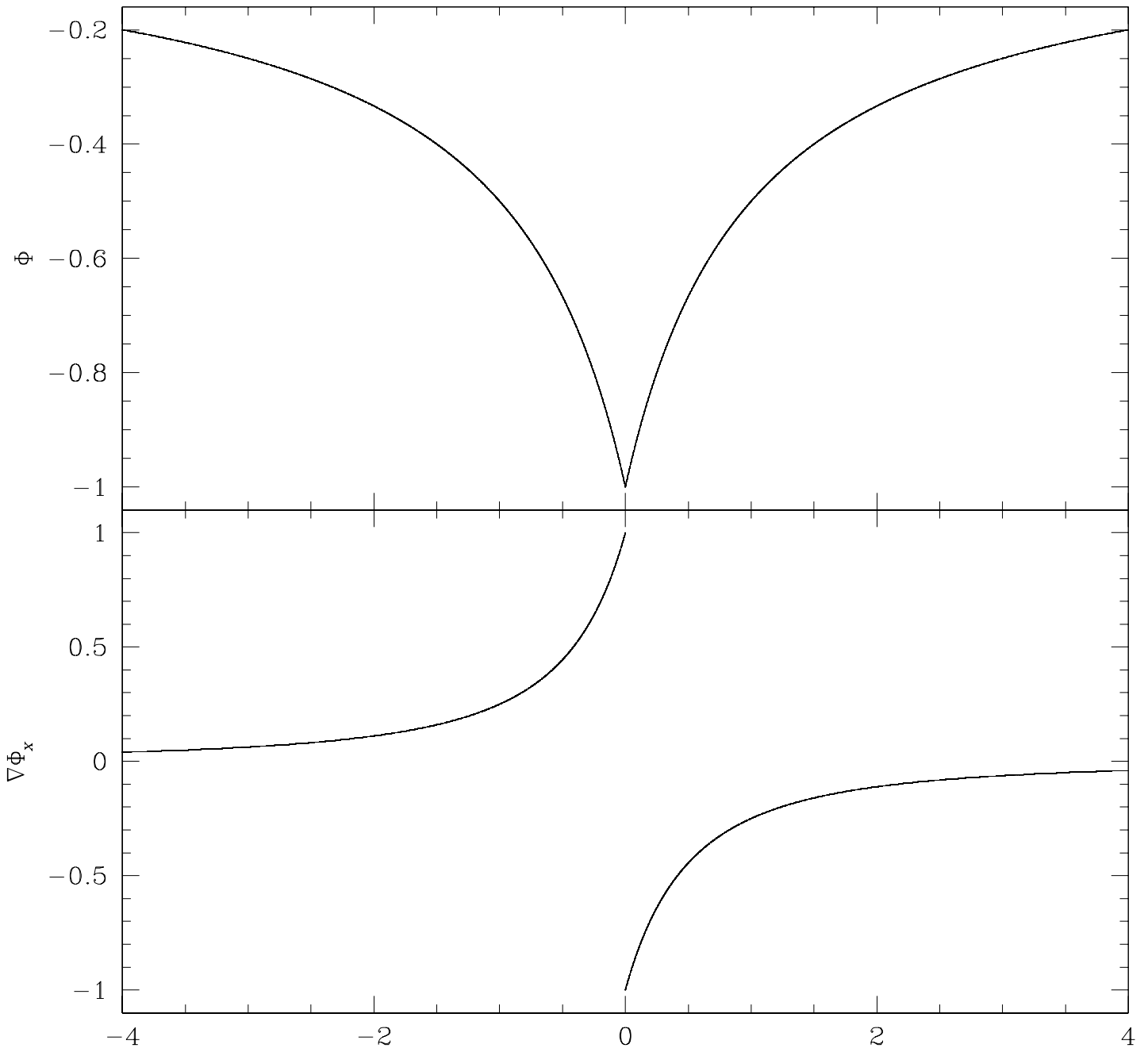
- ▷ general & adaptive
- ▷ problem: f is unknown
- \Rightarrow estimate (ϵ : **softening length**)

$$\Phi(\mathbf{x}_i, t) \approx - \sum_{i \neq j} \frac{G \mu_j}{\sqrt{[\mathbf{x}_i - \mathbf{x}_j(t)]^2 + \epsilon^2}}$$

force softening to

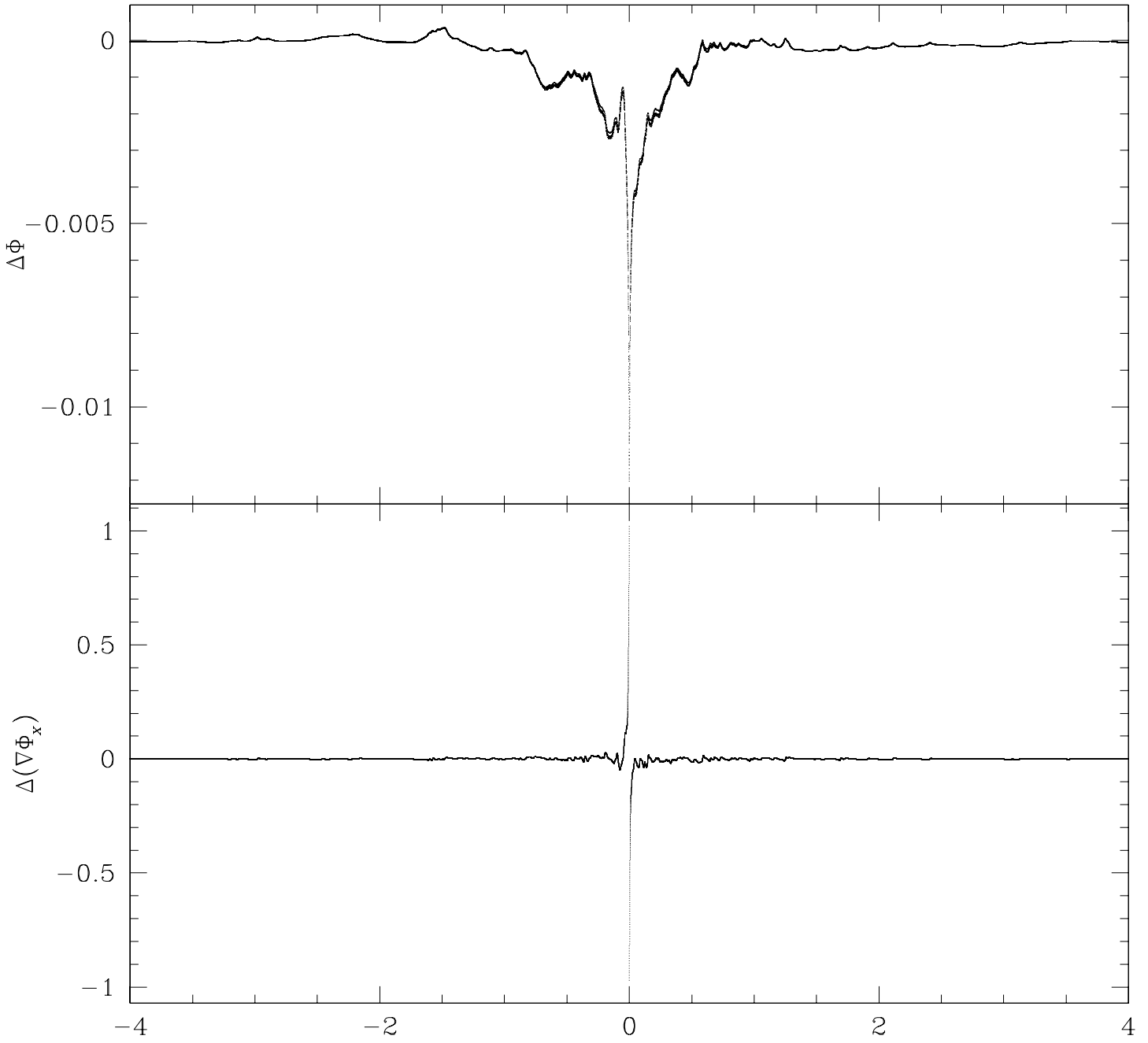
- ▷ optimize force estimate (since f is unknown)
- ▷ suppress (unphysically) close encounters
- \Rightarrow **force-estimation error** (unavoidable)

analytic gravity



true gravity of Hernquist model

estimation error



estimation error with $N = 10^6$

computing the forces

▷ Greens-function approach → Hamiltonian:

$$H = \sum_{i=1}^N \frac{\mu_i}{2} \left[v_i^2 - \sum_{j \neq i} \frac{G \mu_j}{\sqrt{|\mathbf{x}_i - \mathbf{x}_j|^2 + \epsilon^2}} \right]$$

▷ how to evaluate Φ & $\nabla\Phi$?

▷ can tolerate **approximation error** \ll **estimation error**
⇒ use **approximative** methods

1 direct summation (not approximative):

- ▷ slow: $\mathcal{O}(N^2)$ (but: GRAPE)
- ▷ (unnecessarily) accurate
- ▷ used in *collisional N-body* codes

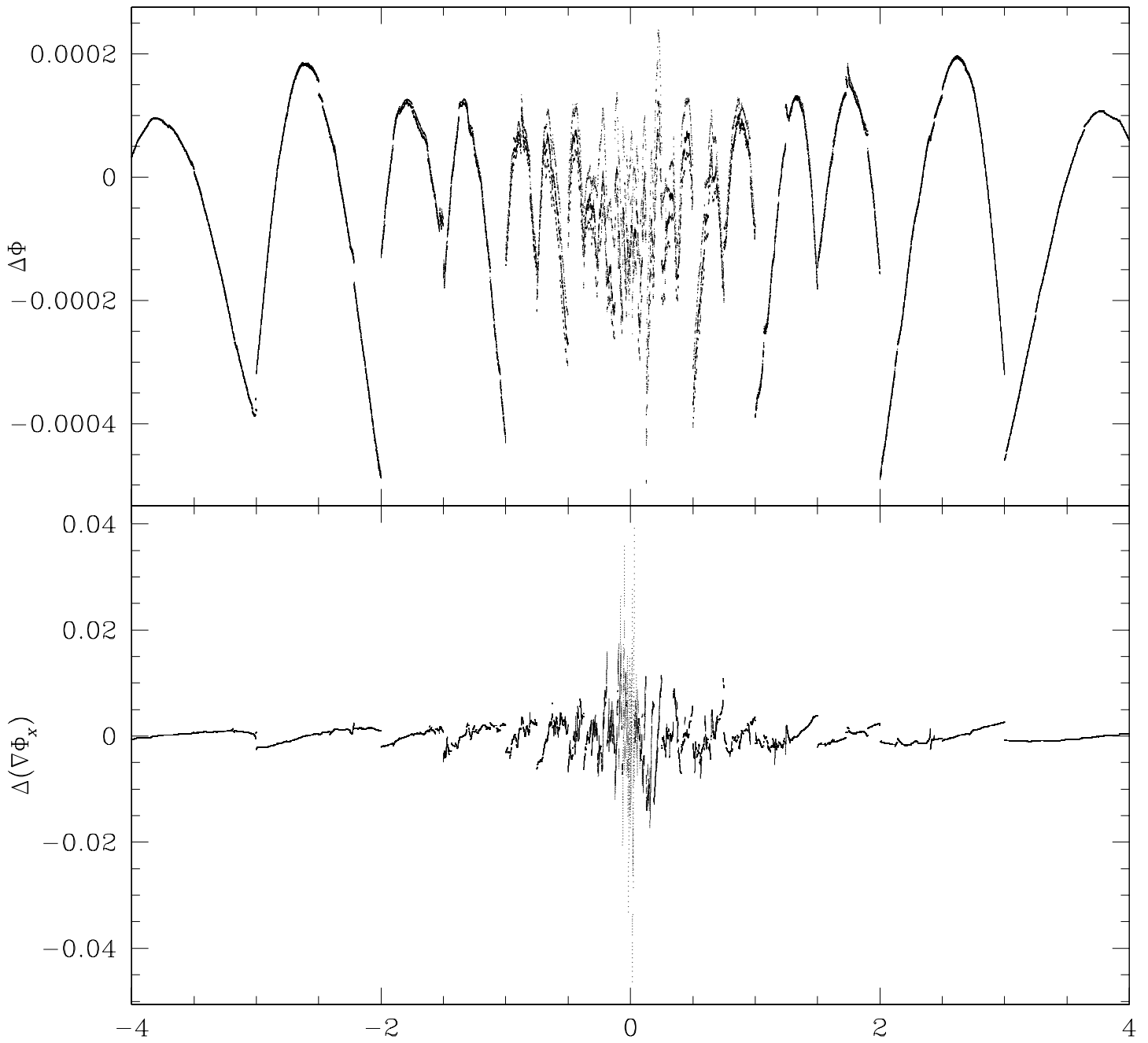
2 Barnes & Hut (1986) **tree code**:

- ▷ use hierarchical tree (usually: oct-tree) ⇒ fully adaptive
- ▷ fast(er): $\mathcal{O}(N \log N)$
- ▷ most common method in astrophysics
- ▷ violates Newton's 3rd law
⇒ total momentum not conserved

3 traditional **fast multipole method (FMM)**:

- ▷ use hierarchy of cartesian grids ⇒ not fully adaptive
- ▷ compute gravity via spherical multipoles & complex Y_{lm}
⇒ numerics complicated & cumbersome
- ▷ formally $\mathcal{O}(N)$, **but**
slower than **tree code** (for astrophysical applications,
see Capuzzo-Colcetta & Miochi, 1998, JCP, **143**, 29)

approximation error



approximation error with $N = 10^6$

details of the tree code

1 preparation phase

1.1 build a hierarchical tree of cubic cells

▷ cost: $\mathcal{O}(N \log N)$

1.2 pre-compute multipole moments etc

2 force computation: 'tree-walk'

▷ for each body: compute force due to root cell

▷ to compute force from cell:

if body is **well-separated** from cell:

compute force from multipole moments

otherwise

sum forces from daughter cells (recursive)

▷ cost: $\mathcal{O}(\log N)$ per body $\Rightarrow \mathcal{O}(N \log N)$

▷ the tree code is wasteful:

forces of neighbours are similar yet independently computed

details of the FMM

here I describe traditional Greengard & Rokhlin (1987) FMM

1 preparation phase

1.1 build a hierarchy of cartesian grids

1.2 pre-compute multipole moments etc (**upward pass**)

2 force computation

2.1 interactions

on each grid level:

- ▷ perform ‘intermediate-field’ interactions:
compute & accumulate multipoles of gravity field

2.2 downward pass

- ▷ pass field-multipoles down the hierarchy
- ▷ compute forces on finest grid
- ▷ theoretical $\mathcal{O}(N)$ not demonstrated in practice
- ▷ not competitive with tree code in low-accuracy regime

details of **falcON**

- ▷ hybrid of tree code & FMM
- ▷ takes the better of each method

1 preparation phase (as for tree code)

1.1 build a hierarchical tree of cubic cells

- ▷ cost: $\mathcal{O}(N \log N)$

1.2 pre-compute multipole moments etc

2 force computation

2.1 interaction phase

- ▷ ‘catch’ all body-body interactions in well-separated node-node interactions:
 - if node-node interaction is executable execute it: accumulate field tensors
 - otherwise split it & continue with child interactions (recursive)
- ▷ cost: (better than) $\mathcal{O}(N)$, dominates

2.2 evaluation phase

- ▷ pass field tensors down the tree
- ▷ compute forces at body positions
- ▷ cost: $\mathcal{O}(N)$

- ▷ ~ 10 times faster than tree code or FMM (at low accuracy)

numerics of **falcON**

Wanted:

$$\Phi(\mathbf{x}_i) = - \sum_{j \neq i} \mu_j g(\mathbf{x}_i - \mathbf{y}_j),$$

Taylor expand g about $\mathbf{R} = \mathbf{x}_0 - \mathbf{y}_0$

$$g(\mathbf{x} - \mathbf{y}) = \sum_{n=0}^p \frac{1}{n!} (\mathbf{x} - \mathbf{y} - \mathbf{R})^{(n)} \odot \nabla^{(n)} g(\mathbf{R}) + \mathcal{R}_p(g),$$

Insert & sum over source cell B

$$\Phi_{\mathbf{B} \rightarrow \mathbf{A}}(\mathbf{x}) = - \sum_{m=0}^p \frac{1}{m!} (\mathbf{x} - \mathbf{x}_0)^{(m)} \odot \mathbf{C}^{m,p} + \mathcal{R}_p(\Phi_{\mathbf{B} \rightarrow \mathbf{A}})$$

$$\mathbf{C}^{m,p} = \sum_{n=0}^{p-m} \frac{(-1)^n}{n!} \nabla^{(n+m)} g(\mathbf{R}) \odot \mathbf{M}_{\mathbf{B}}^n,$$

$$\mathbf{M}_{\mathbf{B}}^n = \sum_{\mathbf{y}_i \in \mathbf{B}} \mu_i (\mathbf{y}_i - \mathbf{y}_0)^{(n)}.$$

(Warren & Salmon 1995: Comp. Phys. Comm, 87, 266)

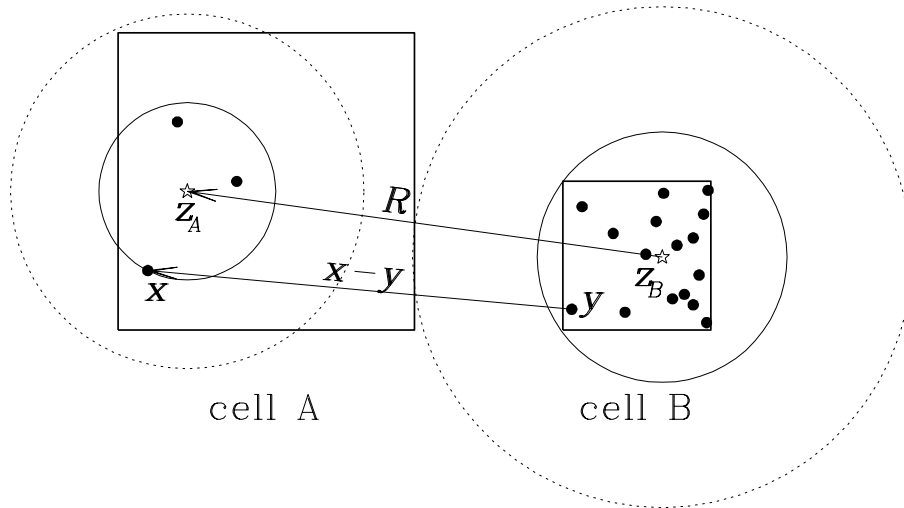
\sum_m : evaluation of gravity, represented by the **field tensors** $\mathbf{C}^{m,p}$, at position \mathbf{x}

\sum_n : interaction between source cell B, represented by the **multipoles** $\mathbf{M}_{\mathbf{B}}^n$, and the sink cell A.

Difference to tree code:

- ▷ **expansion** in \mathbf{x} (tree code: $\mathbf{x} \equiv \mathbf{x}_0$)
- ▷ **mutuality** of interactions

gravity between well-separated nodes



two well-separated cells

If $|R| > r_{A,crit} + r_{B,crit}$ with $r_{crit} = r_{max}/\theta$,

$\Rightarrow |x-y-R| < \theta|R| \forall x \in A, y \in B$ & Taylor series converges

force error of individual interaction:

$$|\nabla \mathcal{R}_p(\Phi_{B \rightarrow A})| \leq \frac{(p+1)\theta^p}{(1-\theta)^2} \frac{M_B}{R^2}$$

$$\propto \frac{\theta^{p+2}}{(1-\theta)^2} r_{B,max}^{d-2} \propto \frac{\theta^{p+2}}{(1-\theta)^2} M_B^{(d-2)/d}$$

▷ standard tree-code & FMM practice: $\theta = \text{const}$

\Rightarrow **relative** error controlled

\Rightarrow **absolute** error increases with M_B

\Rightarrow **total error** dominated by few interactions with large cells

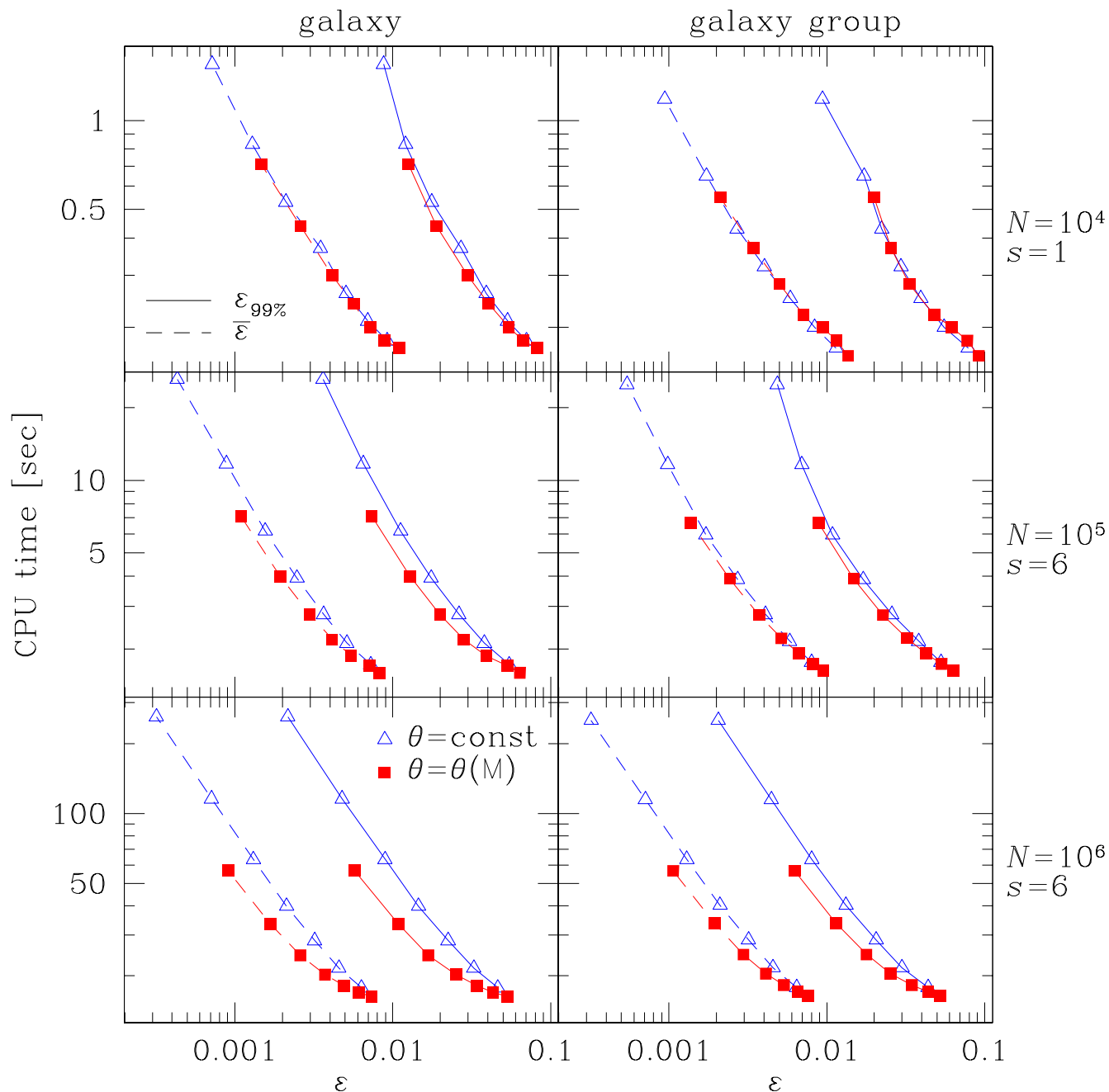
\Rightarrow better:

▷ balance **absolute** individual errors by $\theta = \theta(M)$ with

$$\frac{\theta^{p+2}}{(1-\theta)^2} = \frac{\theta_{min}^{p+2}}{(1-\theta_{min})^2} \left(\frac{M}{M_{tot}} \right)^{(2-d)/d}$$

\Rightarrow reduce total error

accuracy vs. CPU time

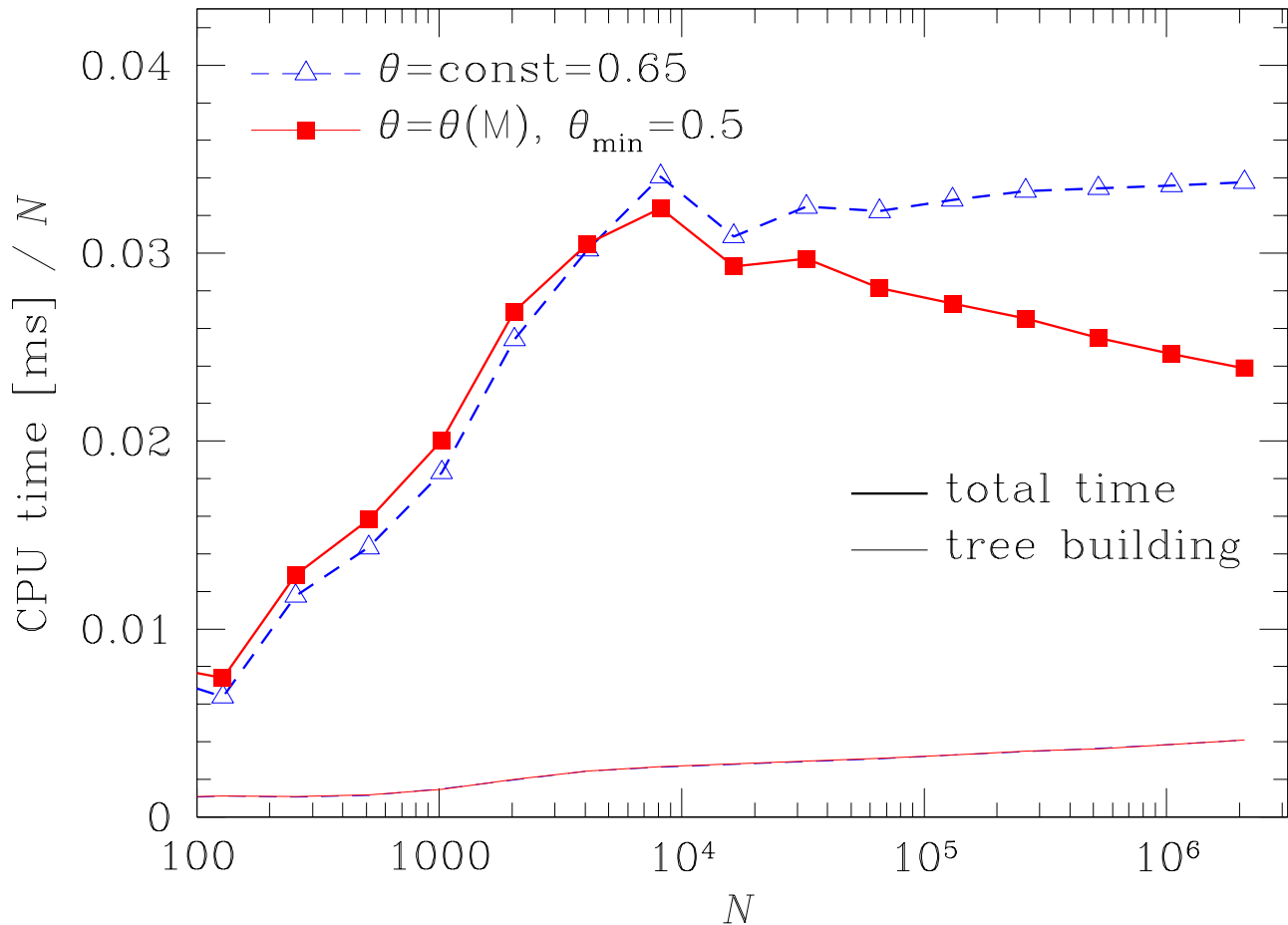


mean (*dashed*) and 99 percentile (*solid*) relative force error

$$\varepsilon \equiv |a_{\text{approx}} - a_{\text{exact}}| / a_{\text{exact}},$$

versus the CPU time (Pentium III/933Mhz in **2001**) for a galaxy (*left*) and a group of galaxies (*right*), sampled with (total) $N = 10^4$ (*top*), $N = 10^5$ (*middle*), or $N = 10^6$ (*bottom*). We used either $\theta = \text{const}$ (*open triangles*) or $\theta = \theta(M)$ (*solid squares*). The symbols along each curve correspond, from left to right, to values for θ or θ_{\min} of 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, and 0.8.

performance



CPU time per body (Pentium III/500Mhz in **2000**) versus N for a galaxy group.

what complexity?

▷ 8-folding $N \Rightarrow N_I \rightarrow 8N_I + N_+$ and thus:

$$\frac{dN_I}{dN} \simeq \frac{N_I \Delta \ln N_I}{N \Delta \ln N} \approx \frac{N_I}{N} + \frac{N_+}{N 8 \ln 8},$$

with solution

$$N_I = c_0 N + \frac{N}{8 \ln 8} \int \frac{N_+}{N^2} dN$$

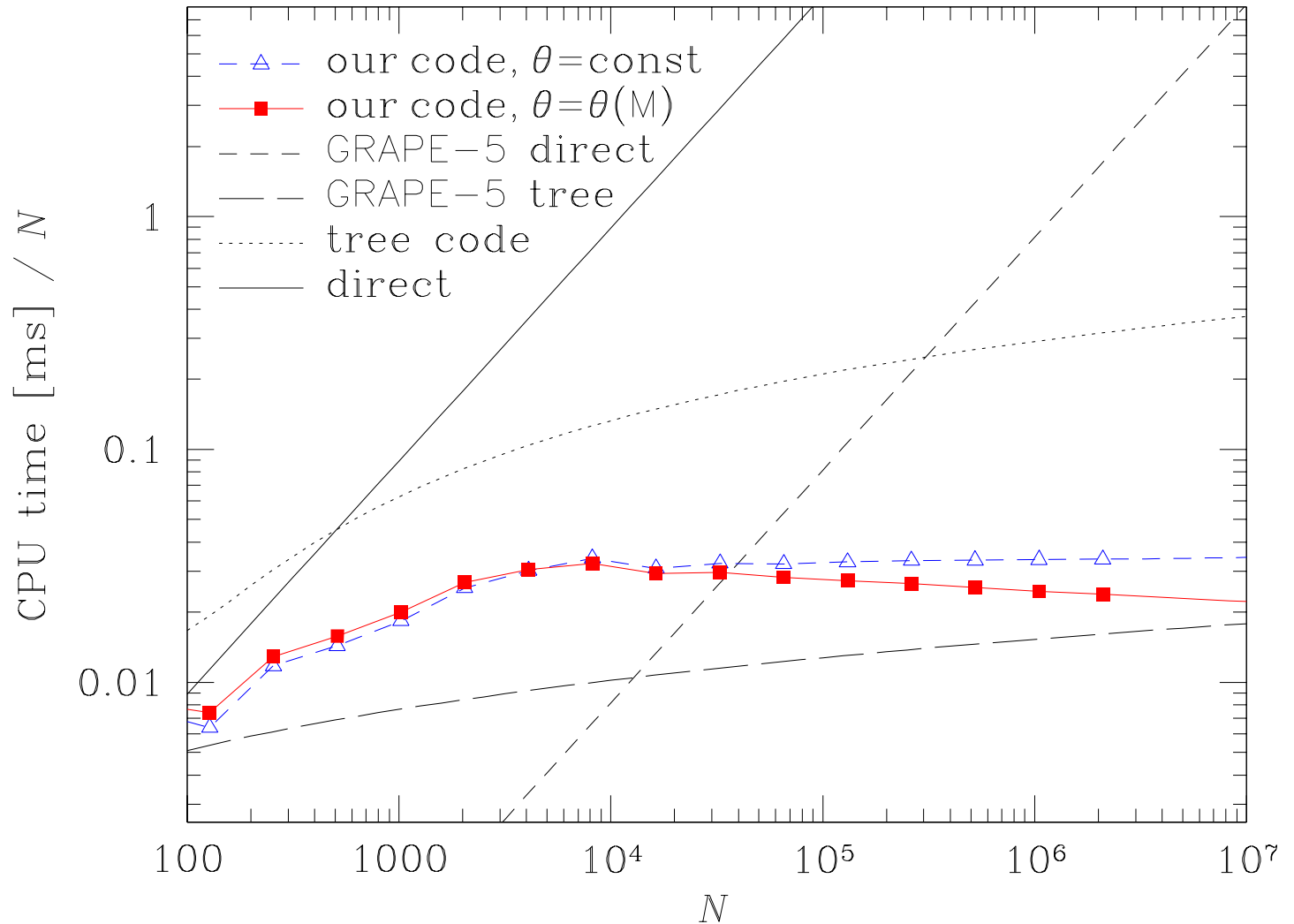
▷ B&H tree code: $N_+ \propto N$

$\Rightarrow N_I \propto N \log N$

▷ Here: $N_+(N)$ grows sub-linear at large N

$\Rightarrow N_I \propto N$

comparison with other methods used in astrophysics



CPU time per body (**2001**) versus **N** for various techniques. Note that there are differences in the hard- & software, stellar system, and accuracy requirements.

by **2003/2004**: **falcON** is ~ 3 times faster, but GRAPE-5 tree not.

comparison with FMM

▷ comparing under same conditions (bodies uniform in a cube)

$$E = \left[\frac{\sum_i (\Phi_{i,\text{direct}} - \Phi_{i,\text{approx}})^2}{\sum_i \Phi_{i,\text{direct}}^2} \right]^{1/2}$$

▷ low-accuracy regime: ~ 10 times faster:

timing results (in seconds):

N	T_{FMM}^a	T_{direct}^a	E^a	T_{falcON}^b	T_{direct}^c	E^b
20000	13.3	233	7.9×10^{-4}	0.97	136	3.7×10^{-4}
50000	27.7	1483	5.2×10^{-4}	2.64	924	3.3×10^{-4}
200000	158	24330	8.4×10^{-4}	10.77	14694	3.4×10^{-4}
500000	268	138380	7.0×10^{-4}	29.42	91134	3.7×10^{-4}
1000000	655	563900	7.1×10^{-4}	58.34	366218	3.5×10^{-4}

^a FMM; data from Table I of Cheng et al. (1999: JCP, 155, 468)

^b **falcON** on a computer identical to that used by Cheng et al.

^c our own implementation of direct summation on the same computer

▷ high-accuracy regime:

falcON cannot compete with FMM

⇒ accuracy & performance depend on both p & θ

▷ FMM: fixed ' θ ', vary p

▷ **falcON**: fixed $p = 3$, vary θ

⇒ high accuracy requires higher order p

summary

- ▷ **falcON** = hybrid of tree code & FMM
- ▷ new features:
 - explicitly exploits mutuality of gravity
 - ⇒ reduces computational effort
 - ⇒ requires novel tree-walking algorithm
 - ⇒ conservation of Newton's 3rd law
 - mass-dependent θ
 - ⇒ error balancing
 - ⇒ reduces cost to **better** than $\mathcal{O}(N)$
- ▷ ~ 10 times faster than tree code or FMM
- ▷ publicly available

more dogmas

- ▷ balance errors
 - ⇒ reduce effort at given accuracy
- ▷ keep algorithm as simple as possible &
as complicated as necessary
 - ⇒ high-order may be unnecessary
- ▷ write efficient code
 - ⇒ avoid cache misses
 - ⇒ data structure design
 - ⇒ write generic code
 - ⇒ do not rely too much on compiler optimization
 - ⇒ template metaprogramming