

# Fast Multipole Method

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**Aiichiro Nakano**

*Collaboratory for Advanced Computing & Simulations*

*Department of Computer Science*

*Department of Physics & Astronomy*

*Department of Chemical Engineering & Materials Science*

*Department of Quantitative & Computational Biology*

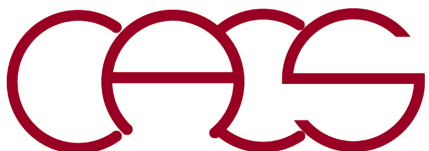
*University of Southern California*

**Email: [anakano@usc.edu](mailto:anakano@usc.edu)**

**Objectives: Space-time multiresolution algorithms**

**> Tree codes: fast multipole method**

**> Multiple time stepping**



# Top 10 Algorithms in History

In putting together this issue of *Computing in Science & Engineering*, we knew three things: it would be difficult to list just 10 algorithms; it would be fun to assemble the authors and read their papers; and, whatever we came up with in the end, it would be controversial. We tried to assemble the 10 algorithms with the greatest influence on the development and practice of science and engineering in the 20th century.

Following is our list (here, the list is in chronological order; however, the articles appear in no particular order):

- Metropolis Algorithm for Monte Carlo  $\pi_{m,n} = \min(\overbrace{\frac{\rho_m}{\rho_n}}^{\text{accept/reject}}, 1) \overbrace{\alpha_{m,n}}^{\text{attempt}}$
- Simplex Method for Linear Programming
- Krylov Subspace Iteration Methods
- The Decompositional Approach to Matrix Computations
- The Fortran Optimizing Compiler
- QR Algorithm for Computing Eigenvalues
- Quicksort Algorithm for Sorting
- Fast Fourier Transform
- Integer Relation Detection
- Fast Multipole Method

PHYS 516

CSCI 596

CSCI 653

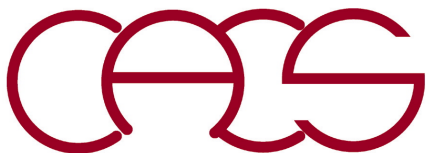
# Data Locality in MD

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- **Spatial locality:** Atoms closer to each other interact more tightly
- **Temporal locality:** Computations performed in consecutive MD time steps are similar



- **Efficient simulation algorithms**
  - > Reduced computational complexity
  - > Better utilization of hardware memory hierarchy (intranode)
  - > Minimized communication overhead (internode)



# Molecular Dynamics: $N$ -Body Problem

- Newton's equations of motion

$$m_i \frac{d^2 \mathbf{r}_i}{dt^2} = - \frac{\partial E_{\text{MD}}(\mathbf{r}^N)}{\partial \mathbf{r}_i} \quad (i = 1, \dots, N)$$

- Reliable interatomic potential

$$E_{\text{MD}} = \sum_{i < j} u_{ij}(r_{ij}) + \sum_{i, j < k} v_{jik}(\mathbf{r}_{ij}, \mathbf{r}_{ik})$$

- $N$ -body problem

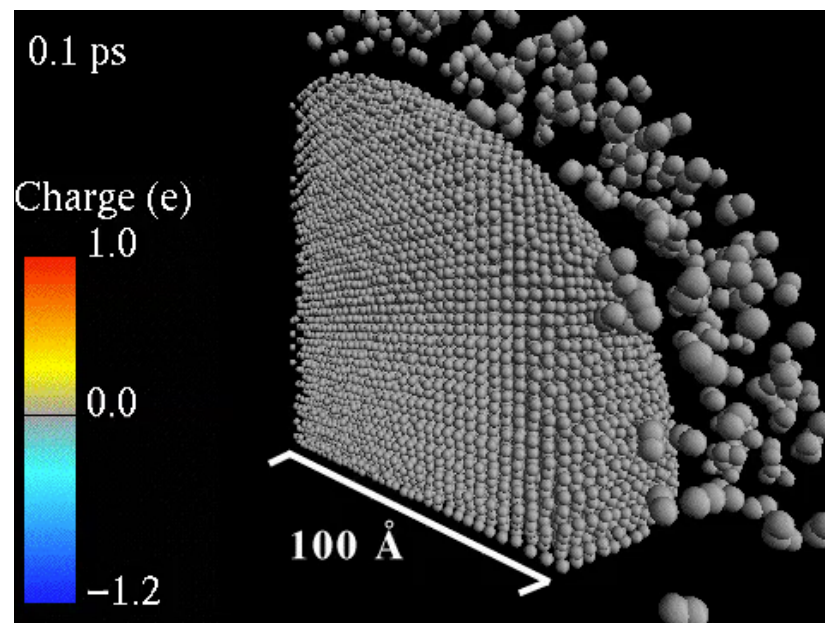
Long-range electrostatic interaction —  $O(N^2)$

$$\text{Evaluate } V_{\text{es}}(\mathbf{x}) = \sum_{i=1}^N \frac{q_i}{|\mathbf{x} - \mathbf{x}_i|} \text{ at } \mathbf{x} = \mathbf{x}_j \ (j = 1, \dots, N)$$

- $O(N)$  space-time multiresolution MD (MRMD) algorithm

1. Fast multipole method (FMM) [Greengard & Rokhlin, '87]

2. Symplectic multiple time stepping (MTS) [Tuckerman *et al.*, '92]

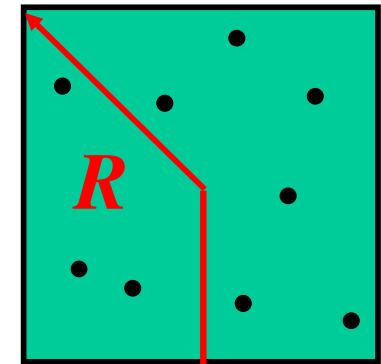


# Clustering in the Fast Multipole Method

- Encapsulate far-field information in terms of multipoles at the source

$$V(\mathbf{x}) = \sum_{i=1}^N \frac{q_i}{|\mathbf{x} - \mathbf{x}_i|}$$

$$= \sum_{l=0}^{\infty} \sum_{m=-l}^l \left\{ \sum_{i=1}^N q_i r_i^l Y_l^{*m}(\theta_i, \phi_i) \right\} \frac{Y_l^m(\theta, \phi)}{r^{l+1}}$$

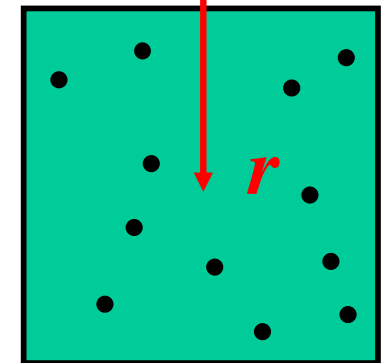


- Well-defined error bound

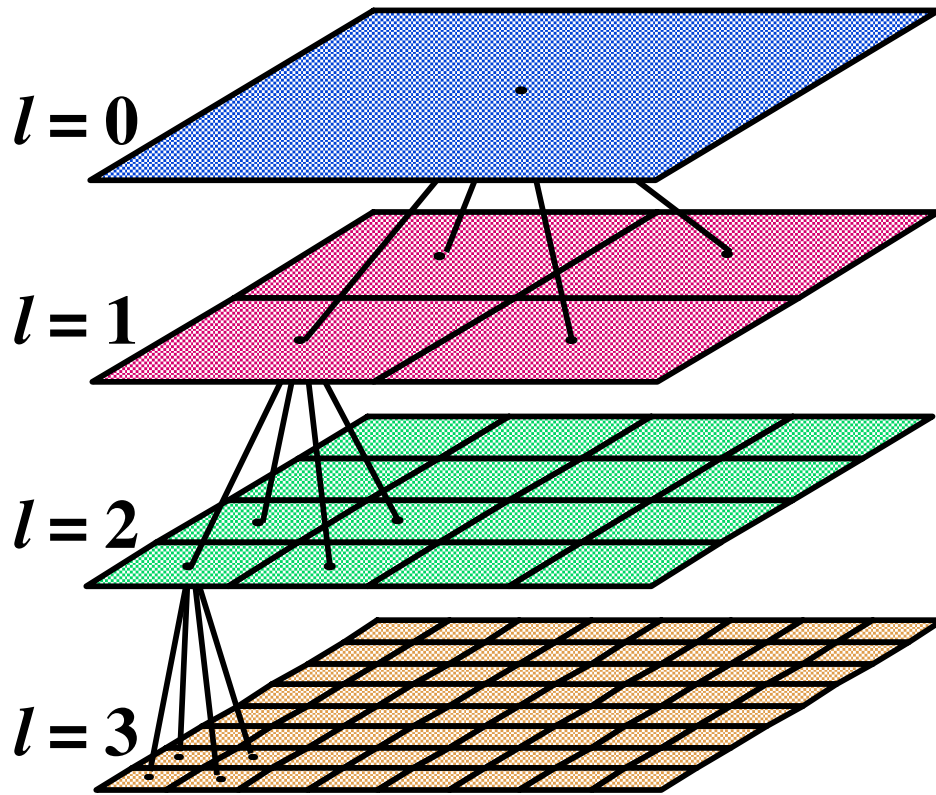
$$\text{Error} \leq \frac{\sum_{i=1}^N |q_i|}{r - R} \left( \frac{R}{r} \right)^{p+1} \quad p: l\text{-sum truncation}$$

- Local expansion at the destination

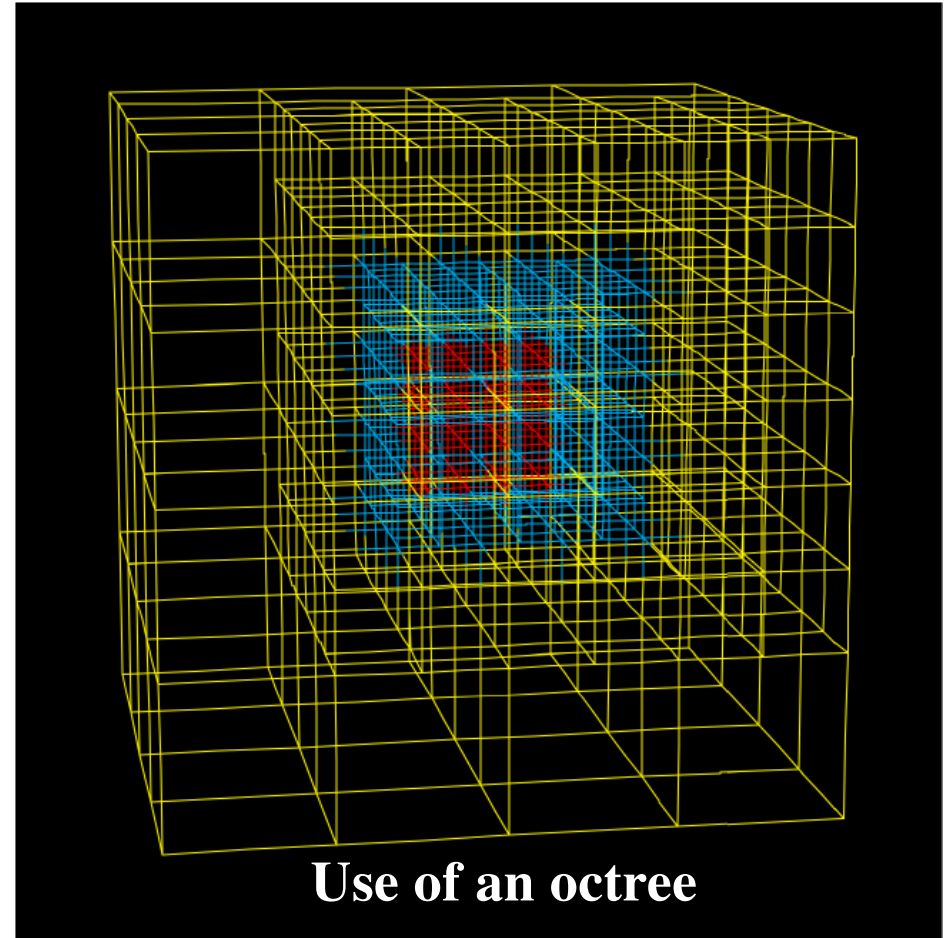
$$V(\mathbf{x}) = \sum_{l=0}^{\infty} \sum_{m=-l}^l \left\{ \sum_{i=1}^N \frac{q_i Y_l^m(\theta_i, \phi_i)}{r_i^{l+1}} \right\} r^l Y_l^{*m}(\theta, \phi)$$



# Hierarchical Abstraction



2D example



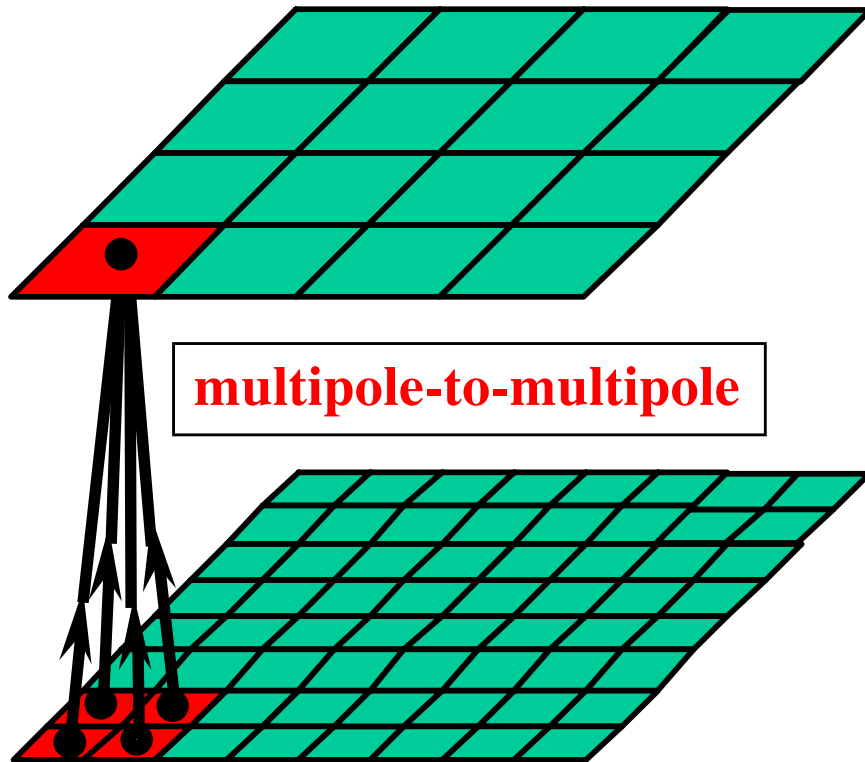
Use of an octree

- Larger clusters for longer distances to keep the error constant
- Recursively subdivide the simulation box to form an octree in 3D or quadtree in 2D

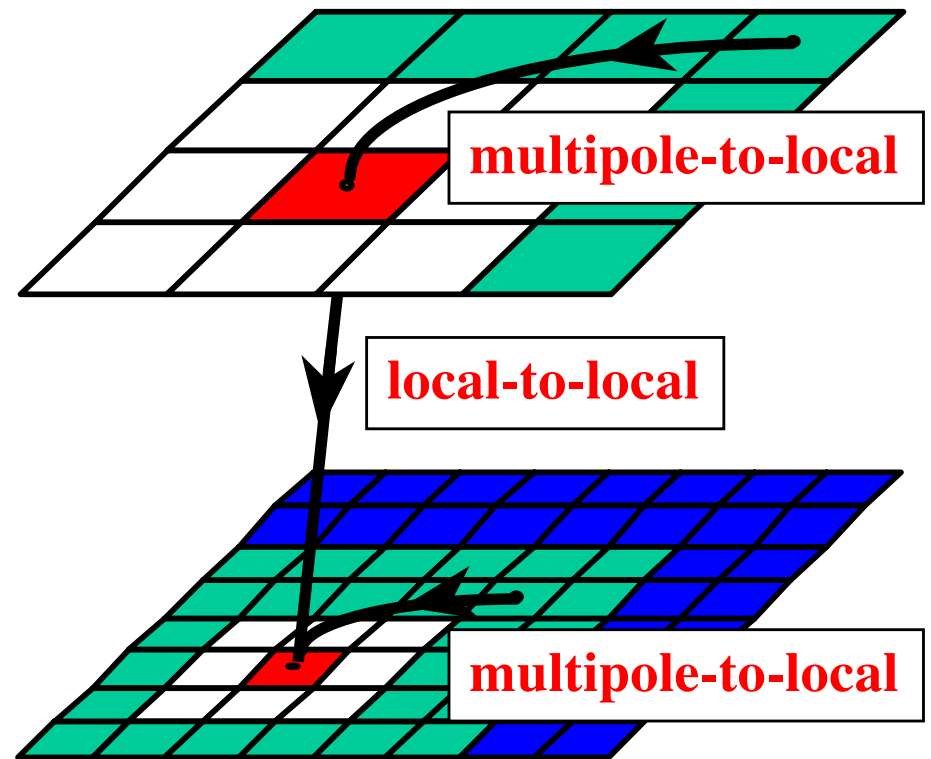


# $O(N)$ Algorithm

Upward Pass



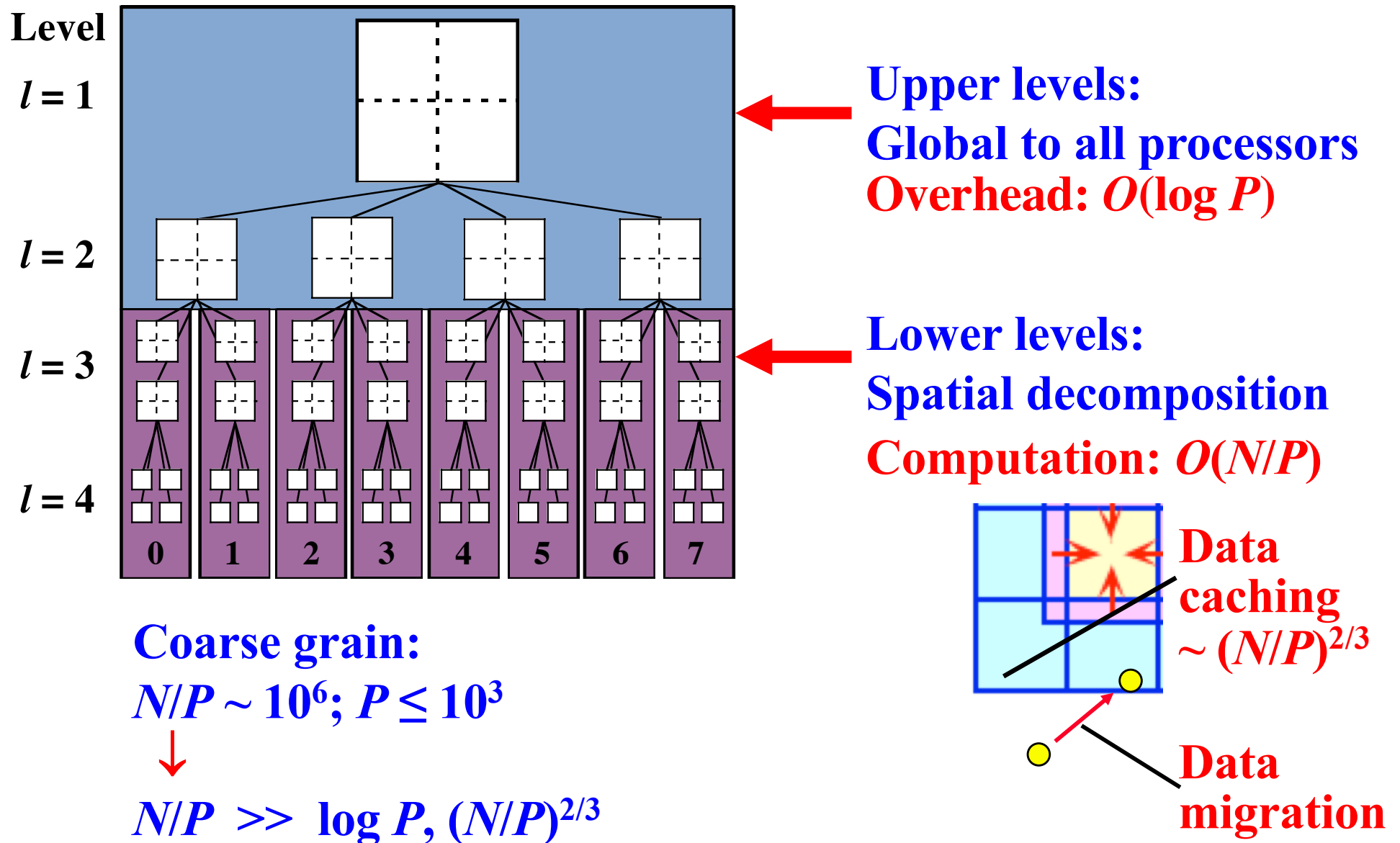
Downward Pass



1. **Upward pass** computes multipoles for all cells: M-to-M translation
2. **Downward pass** translates multipoles to local terms for all cells
  - Constant (189 in 3D) **interactive (cousin) cells** per destination cell contribute to M-to-L translation Within parent's neighbor but not my neighbor
  - **Inheritance** from the parent cell: L-to-L translation (& **delegation**)
3. **Direct interactions** for the nearest-neighbor leaf cells

See lecture notes for the MM, ML & LL formula in 2D

# Parallel Implementation of FMM





# FMMP Code Dissemination



**CPC**

Program Library

*Computer programs in physics and physical chemistry*

[\[ADRX Licence\]](#) | [Download ADRX](#) | [E-mail ADRX](#) | (16 Kbytes)

**SCALABLE AND PORTABLE IMPLEMENTATION OF THE FAST  
MULTIPOLE METHOD ON PARALLEL COMPUTERS.**

**S. OGATA, T.J. CAMPBELL, R.K. KALIA, A. NAKANO, P. VASHISHTA, S.  
VEMPARALA.**

## **PROGRAM SUMMARY**

**Title of program:** FMMP

**Catalogue identifier:** ADRX

**Journal reference:** Comput. Phys. Commun. 153(2003)445 [[Article index](#)]

**Distribution format:** tar gzip file

**Operating system:** LINUX with MPICH, IBM SP, SGI Origin

**Number of lines in distributed program, including test data, etc:** 3179

**Keywords:** Fast multipole method, Parallel computation, Stress calculation, Periodic boundary condition, Coulomb interaction, Electrostatics.

**Programming language used:** Fortran, C

**Computer:** IBM SP3 .

## **Nature of problem:**

Parallel computations of Coulomb potentials, forces, and stress tensors for a collection of charged particles.

## **Method of solution:**

The fast multipole method.

## **Typical running time:**

Proportional to the number of charged points.

## **Unusual features:**

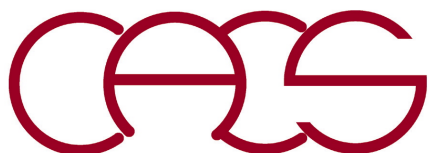
None

## **Additional comments:**

The code has been parallelized using MPI Standard.

<http://www.cpc.cs.qub.ac.uk>

Catalogue ID: ADRX



# $O(N)$ Psuedo-charge Method for Calculating Stresses in Coulombic Systems

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**Microscopic stress tensor:**  $\vec{\pi}_i = \sum_{j \neq i} \vec{r}_{ij} \vec{f}_{ij}^T = q_i \sum_{j \neq i} q_j \frac{\vec{r}_{ij} \vec{r}_{ij}^T}{r_{ij}^3}$

**Attach information of particle position to the particle charge:**

$$\vec{C}(\vec{r}_i, \vec{k}) = - \sum_{j \neq i} \frac{\vec{r}_{ij}}{r_{ij}^3} q_j \exp(i\vec{k} \cdot \vec{r}_j)$$

**Stress tensor is obtained by numerical differentiation:**

$$\frac{\partial}{\partial k_\alpha} \{C_\beta(\vec{r}_i, \vec{k}) \exp(-i\vec{k} \cdot \vec{r}_i)\} \xrightarrow{k_\alpha \rightarrow 0} = i q_j \sum_{j \neq i} \frac{r_{ij}^\alpha r_{ij}^\beta}{r_{ij}^3}$$

**No need for multipole translation operators for the stress tensor, which are unknown**