

# Moore's Law

- Every 18 months, the speed of your computer is doubled
- Every 18 months, the memory on your computer is doubled
- At the same time, the cost of your computer goes down - not quite exponentially, because the box does not become much cheaper!
- A good number to look at

$$R_{1970} = \frac{\textit{Cost of CPU time}}{\textit{Cost of human time}}$$

- 1970 is the year
- Different CPUs, different humans, etc.

# Observation

- $R_{1945} \gg 1000$
- $R_{1960} \gg 100$
- $R_{1970} \gg 10$
- $R_{1980} \sim 1$
- $R_{2000} \ll 0.01$
- Unlike men, not all CPUs are created equal!  
But then, most CPUs do not vote...
- The thing is not slowing down, though eventually . . .
- What should we be doing as applied mathematicians, numerical analysts, etc.?

# Consequences

- Ticket reservations
- Phone systems
- Tactical bombing
- Experimental science
- Manufacturing
- . . . .

## Missing from the list

- Philosophy
- Theater
- Politics
- Dealing with teen-age children
- Mathematics
- Numerical simulation of physical phenomena (???)

# Subject of the Talk

- Neither the numerical algorithms nor the paradigms for their application have kept pace with the developments of the computer hardware
- There are identifiable reasons for this, and to some extent, remedies can be devised and implemented
- In several environments, the results have been spectacular
- The usual message of an extremist: we are the future, with us or against us, victor or victim
- A somewhat different message for a mathematician

# Structure of the Talk

- Changing paradigm in the numerical use of computers
- Interaction of Moore's law with numerical algorithms
- Characteristics of a modern numerical algorithm
- Example: Gravitational  $n$ -body problem
- Pontification

# Paradigm as of 1945

- Critical mission (Manhattan project, for example)
- Willingness to expend human time on programming (ouch!), debugging of the numerical scheme, interpretation
- Limited computer resources: only small-scale problems can be solved
- Extremely uncomfortable programming environment
- Air of heroism and desperation
- No difference between theoretical numerical analysts and practitioners
- Numerical approaches appropriate to small-scale problems
- Numerical algorithms usually written from scratch

# Paradigm as of 1970

- Mission not necessarily critical (oil exploration, NACA airfoils, more involved aerodynamics, civil and mechanical engineering, rocket fuel stoichiometry, . . .)
- Willingness to expend human time on programming (still pretty uncomfortable), interpretation
- Much improved computer capabilities; CPU time still quite expensive, but the flop rate is much higher; one can try running things at night
- The air much less heroic; most applications in non-desperate environments
- Numerical algorithms appropriate to small-scale problems
- Most numerical codes are written from scratch



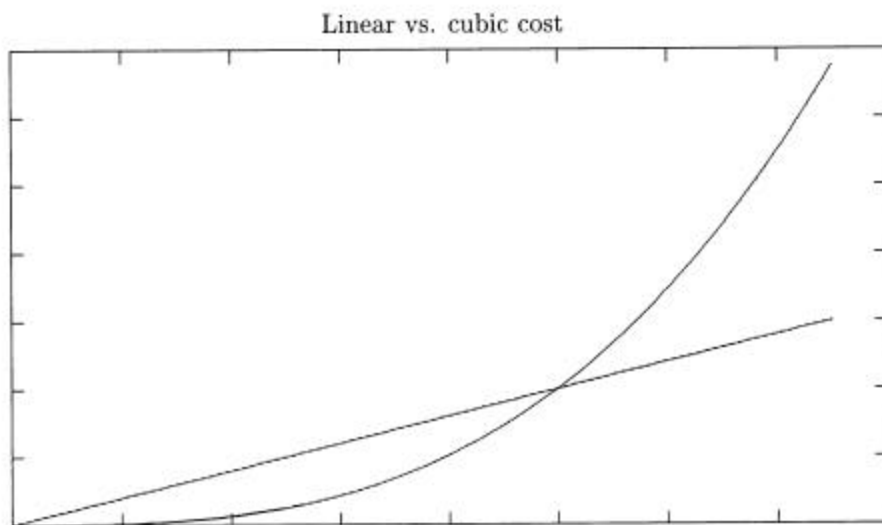
# Paradigm as of 2000

- Mission usually not critical: computer games, medical imaging, design of fishing rods, Boeing-767's . . .
- Limited willingness to expend human time on programming (could be fun, though!), interpretation. . . and most interpreters are not named Teller, Ulam, or Fermi. . .
- Very much improved computer capabilities; CPU time dirt cheap, and flop rate is about to become gigaflop rate
- Air not heroic at all; lots of applications, and most in non-desperate environments
- Numerical algorithms appropriate to small-scale problems
- Most numerical codes are written from scratch

## The Purpose of a Modern Numerical Algorithm

- Produce engineering (physical, biochemical, etc.) results with a minimum expenditure of *human* time
- CPU time is irrelevant *as long as it is affordable* (!!!)
- Note to the algorithm designer: torpedoes should not be aimed at the present location of the ship!

Illustration: Algorithms with CPU time estimates  $O(n^3)$ ,  $O(n \cdot \log(n))$

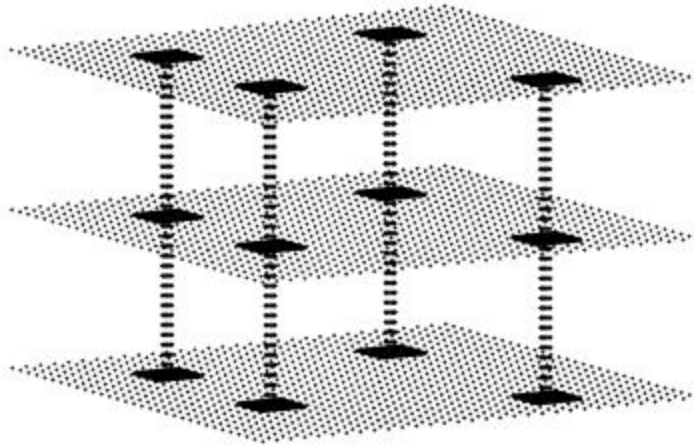


- To a large extent, the choice of the algorithm is determined by the power of one's computer (!!)

# What *do* We Want from a Numerical Algorithm?

- Speed, in the asymptotic sense
- Adaptivity
- Robustness
- Rapid convergence and controlled accuracy: fallacy of the “engineering accuracy” argument; high cost of low precision
- Surprise: adaptivity implies controlled condition numbers, implies (more or less) integral vs. differential equations, implies fast algorithms
- Related surprise: in order to be efficient (or even simply useful), certain algorithms have to be fairly complicated (think about modern cars)

## Numerical $N$ -Body Problem



The calculation of all pairwise interactions in a system of  $N$  particles requires  $O(N^2)$  work.

## Particle Simulations

- ▷ Molecular Dynamics
- ▷ Fluid Dynamics
- ▷ Plasma Physics
- ▷ Dislocations and Plastic Deformation
- ▷ Astrophysics
- ▷ . . . .

## Integral Equations

- ▷ Capacitance calculations
- ▷ Dielectric interface problems
- ▷ Electrodeposition
- ▷ Elasticity
- ▷ Potential flow
- ▷ Incompressible Fluid Dynamics
- ▷ . . .

## Alternative Approaches

- ▷ Field Methods  
(Based on Fast Solvers, FFT)
- ▷ Hierarchical Methods  
(Based on clustering at varying spatial scales)
- ▷ Wavelet, SVD Methods  
(Based on compression of operators)

## Critical Issues

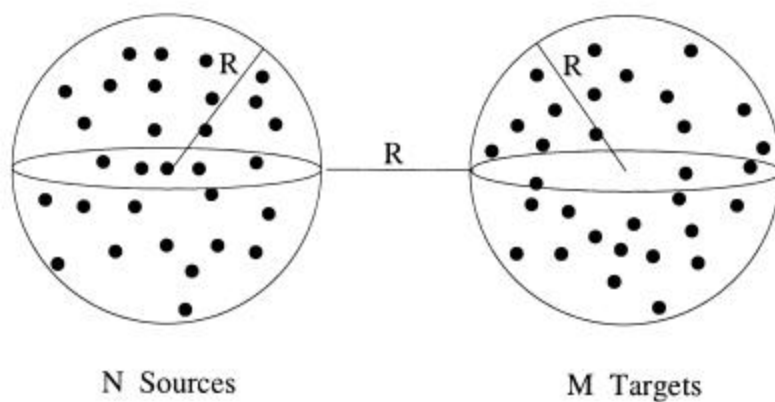
- speed
- adaptivity
- ease of use



## Overview of the Remainder of the Talk

- Analytic Preliminaries
- A simple  $O(N \log N)$  algorithm
- The original FMM
- The modern FMM
- Pontification

## A simple example

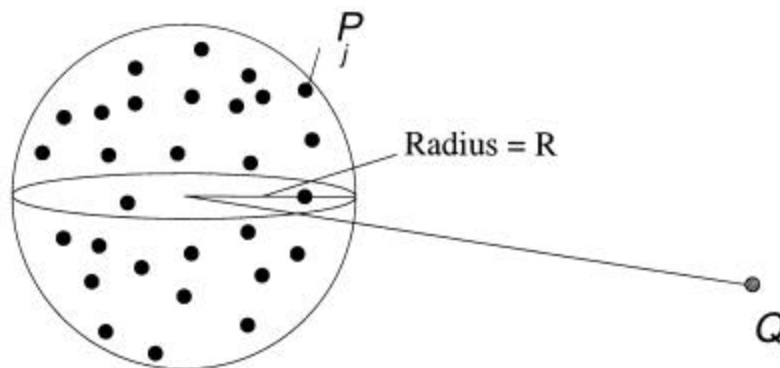


$$V(Q_i) = \sum_{j=1}^N \frac{q_j}{\|Q_i - P_j\|}$$

Direct evaluation requires  $O(NM)$  work.

Newton knew how to fix it...

## Multipole expansion



$$V(Q) = V(r, \theta, \phi) \approx \sum_{n=0}^p \sum_{m=-n}^n \frac{M_n^m Y_n^m(\theta, \phi)}{r^{n+1}},$$

with multipole moments

$$M_n^m = \sum_{j=1}^N q_j Y_n^{-m}(\theta_j, \phi_j) r_j^n, \quad P_j = (r_j, \theta_j, \phi_j)$$

The error in the multipole approximation decays like  $(R/|Q|)^{p+1}$ .

For our simple example,  $R/|Q| < 1/2$ , so that setting  $p = \log_2(\frac{1}{\epsilon})$  yields a precision of  $\epsilon$ .

## Using multipole expansions

- ▷ Evaluate multipole coefficients  $M_n^m$  for  $n = 0, \dots, p$ .
- ▷ Evaluate expansion at target points  $Q_j$ , for  $j = 1, \dots, M$ .
- ▷ Total operation count:  $p^2 \cdot (N + M) = (N + M) \cdot \log^2(\frac{1}{\epsilon})$

## The Fast Multipole Method (FMM)

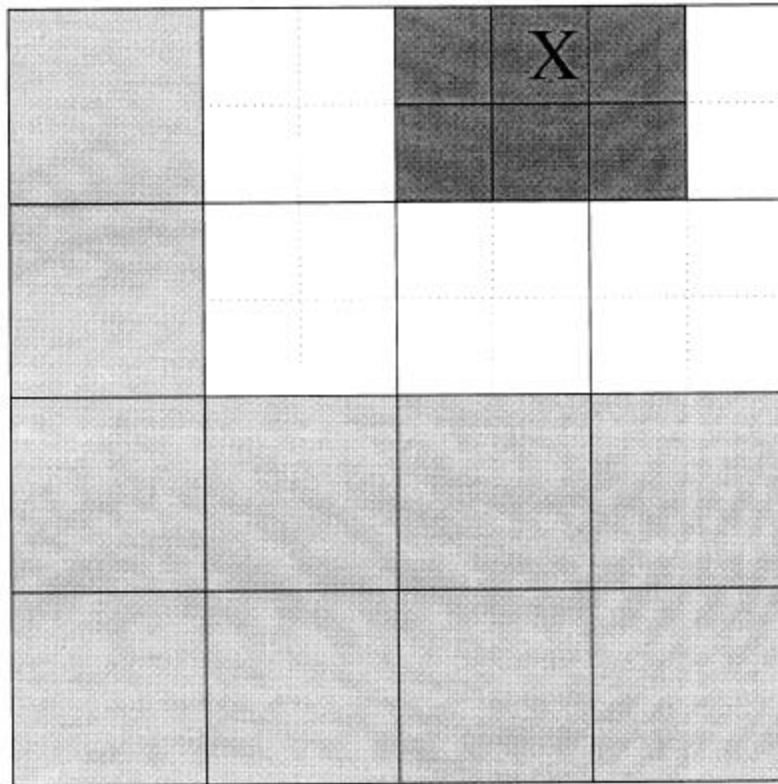
For more general distributions of sources and targets, FMM couples previous analysis with a divide & conquer strategy.

- ▷ Clustering at various spatial length scales
- ▷ Interactions with distant clusters computed by means of multipole expansions
- ▷ Interactions with nearby particles computed directly
- ▷ Fully adaptive algorithm  
Performance essentially independent of particle distribution

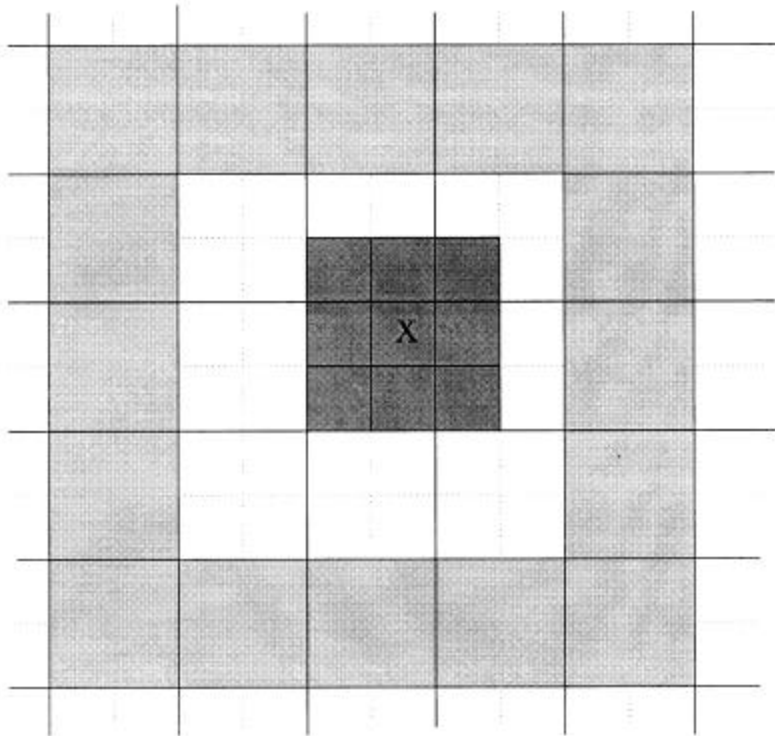
**Step 1:  $N \log N$  Scheme**

		X	

## Step 2: $N \log N$ Scheme



## Step M: $N \log N$ Scheme

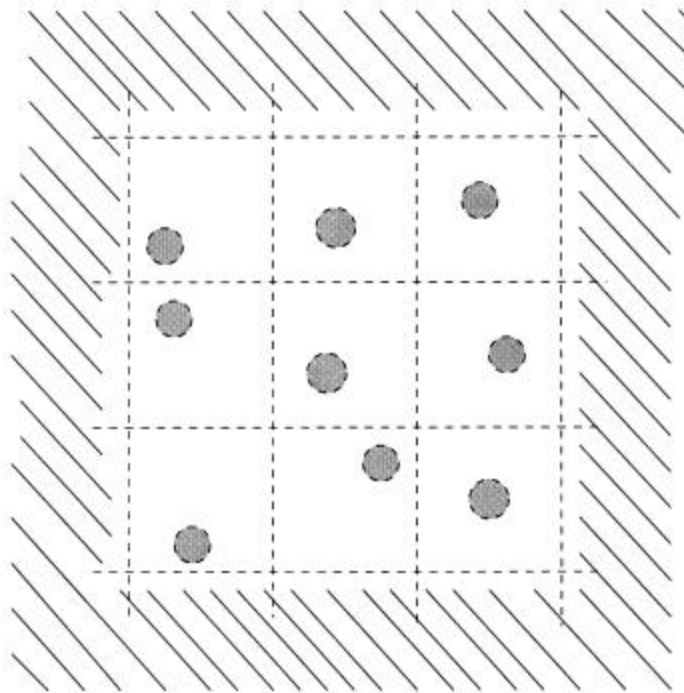




## Final Step: $N \log N$ Scheme

Terminate procedure after  $O(\log_8(N))$  steps.

Total operation count:  $O(N \cdot \log_8 N \cdot p^2)$ ,  
where  $p = \log_c(\frac{1}{\epsilon})$  and  $c = 3/\sqrt{3} \approx 1.73$ .



**Nearest neighbors:**  $O(N)$  work.

## Optimization of constants

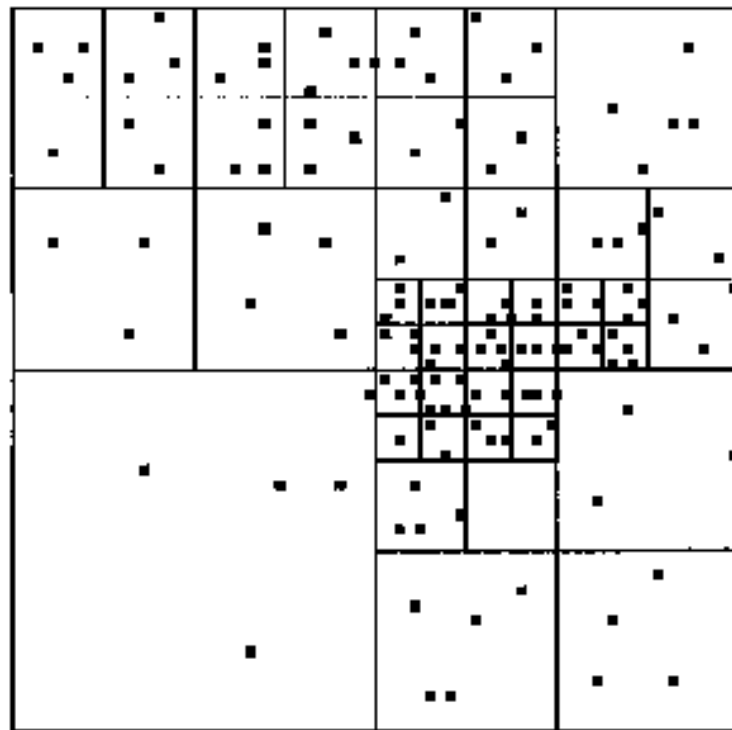
Assume that the distribution is uniform and let  $s$  be the number of particles per box at the finest level.

Multipole expansion work =  $189 N p^2 \log_8(N/s)$ .  
Nearest neighbor work =  $O(27Ns)$ .

Optimal value for  $s$  is

$$s \approx p^2.$$

## Adaptive algorithm

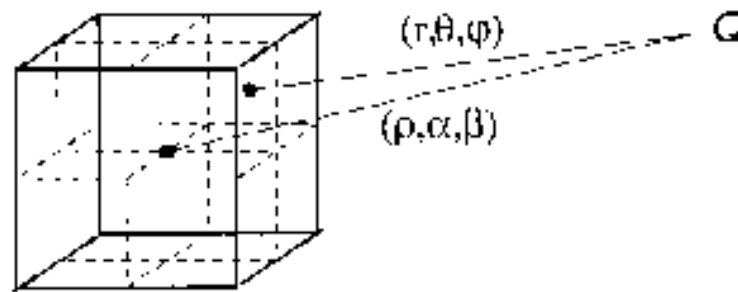


Final stage of subdivision process.

## The order $O(N)$ algorithm

- ▷ Several analytical prerequisites
- ▷ Richer structure
- ▷ Many possible variants

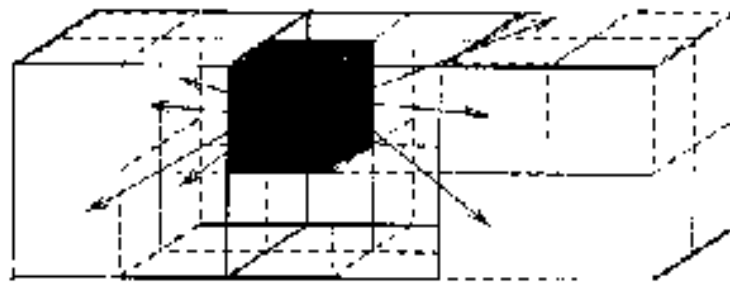
## Translation of multipole expansion



$$\sum_{n=0}^p \sum_{m=-n}^n \frac{M_n^m Y_n^m(\theta, \phi)}{\rho^{n+1}} \rightarrow \sum_{n=0}^p \sum_{m=-n}^n \frac{N_n^m Y_n^m(\alpha, \beta)}{\rho^{n+1}}$$

Cost:  $O(p^4)$  work

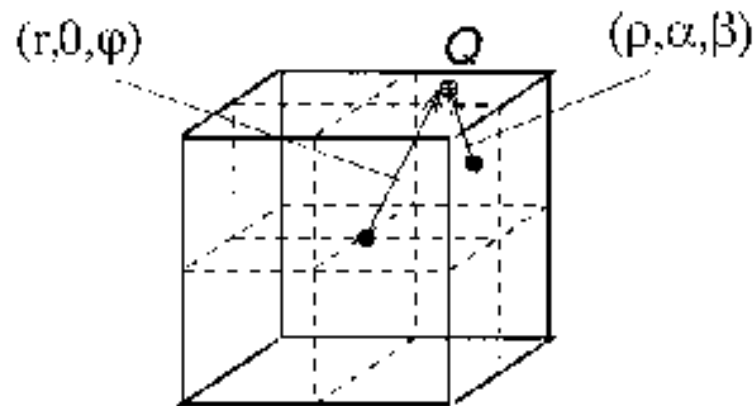
## Construction of local expansion



$$\sum_{n=0}^p \sum_{m=-n}^n \frac{M_n^m Y_n^m(\theta, \phi)}{r^{n+1}} \rightarrow \sum_{n=0}^p \sum_{m=-n}^n L_n^m Y_n^m(\alpha, \beta) \rho^n$$

Cost:  $O(p^4)$  work

## Translation of local expansion



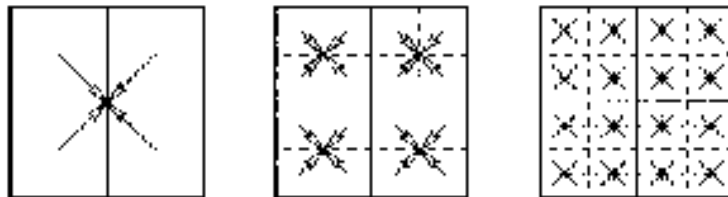
$$\sum_{n=0}^p \sum_{m=-n}^n L_n^m Y_n^m(\theta, \phi) r^n \rightarrow \sum_{n=0}^p \sum_{m=-n}^n O_n^m Y_n^m(\alpha, \beta) \rho^n$$

Cost:  $O(p^4)$  work

## Complexity analysis

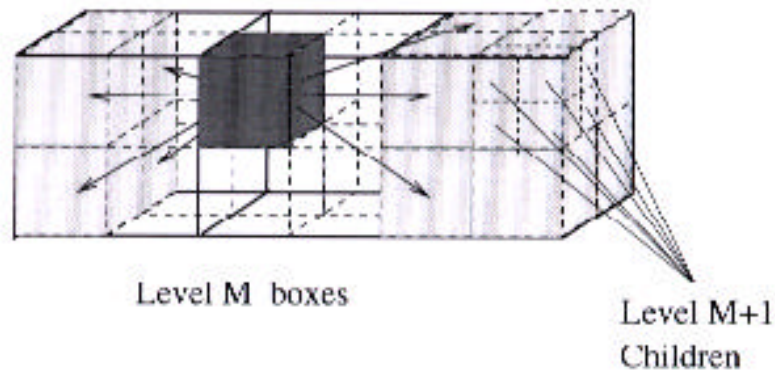
Why  $N \log N$ ?

- ▷ Forming multipole expansions.
- ▷ Evaluating multipole expansions.





## Capture far field in local expansions



- ▷ Use multipole to local translations
- ▷ Use translation of local expansion to transmit information to children

## The order $N$ algorithm

### **Upward Pass**

- ▷ Form multipole expansions at finest level (from source positions and strengths)
- ▷ Form multipole expansions at coarser levels by merging

### **Downward Pass**

- ▷ Account for interactions at each level by conversion lemma
- ▷ Transmit information to finer levels by shifting lemma

## Total operation count

$$189 \frac{N}{s} p^4 + 2 N p^2 + 54 N s$$

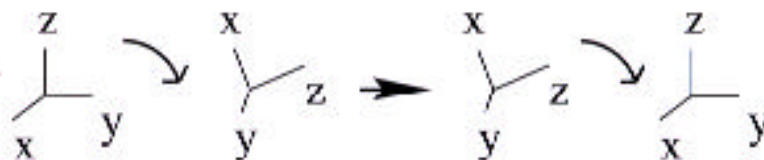
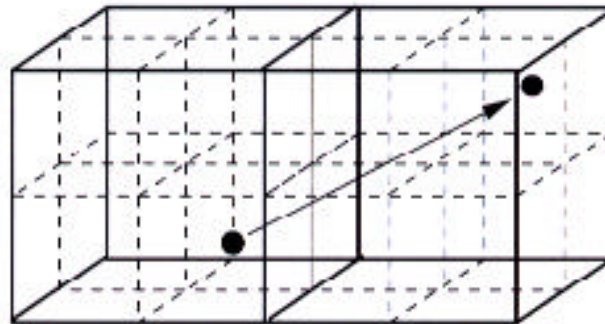
Setting  $s = 1.5 p^2$ , the total operation count is

$$200 N p^2.$$

Recall that the optimal  $N \log N$  scheme required

$$189 N p^2 \left(1 + \log \frac{N}{7p^2}\right) \text{ operations.}$$

## Fast translations I: Rotation



$3p^3$  work is required for each shift, so the total operation count is

$$189 \frac{N}{s} 3p^3 + 2Np^2 + 54Ns.$$

Setting  $s = 3p^{3/2}$ , the total operation count is

$$351 N p^{3/2} + 2Np^2.$$

## Diagonal translation: (G & R, 1988)

- ▷ Based on observation that translations are nearly convolutional
- ▷ Diagonalized by Fourier Transform
- ▷ Numerically unstable
- ▷ Can be stabilized by substructuring (Board et al. 1995)

Operation count

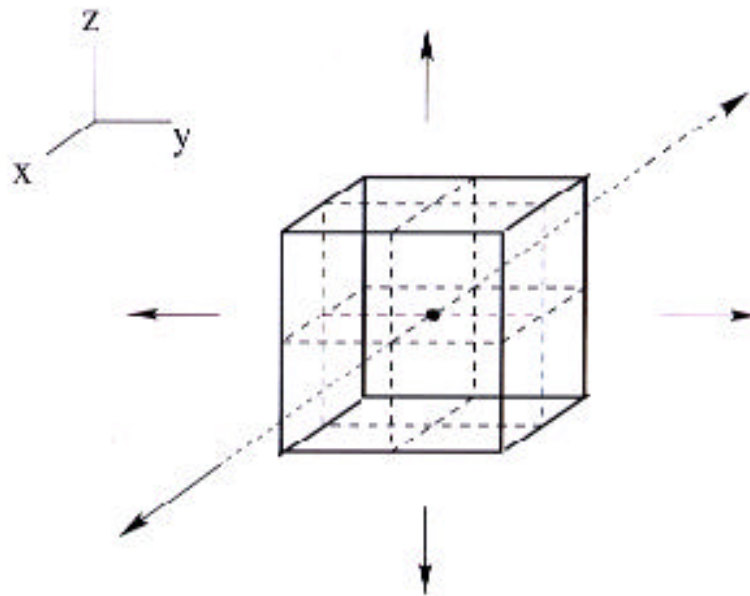
$$189 \frac{N}{s} (2p)^2 + 2 N p^2 + 54 N s + \frac{N}{s} p^2 \log p.$$

Setting  $s = 1.5 p$ , the total operation count is

$$550 N p + 2 N p^2 + \frac{2}{3} N p \log p.$$

## The new FMM

- ▷ 2D scheme : Hrycak and Rokhlin (1995)
- ▷ Based on expansion in plane waves
- ▷ Requires additional analytical machinery





## Exponential representation (+ z)

$$\frac{1}{r} = \frac{1}{2\pi} \int_0^\infty e^{-\lambda z} \int_0^{2\pi} e^{i\lambda(x \cos \alpha + y \sin \alpha)} d\alpha d\lambda.$$

- ▷ Discretization of  $\alpha$  integral: trapezoidal rule
- ▷ Discretization of  $\lambda$  integral: Laguerre or generalized Gaussian quadrature (Yarvin & Rokhlin, 1996)

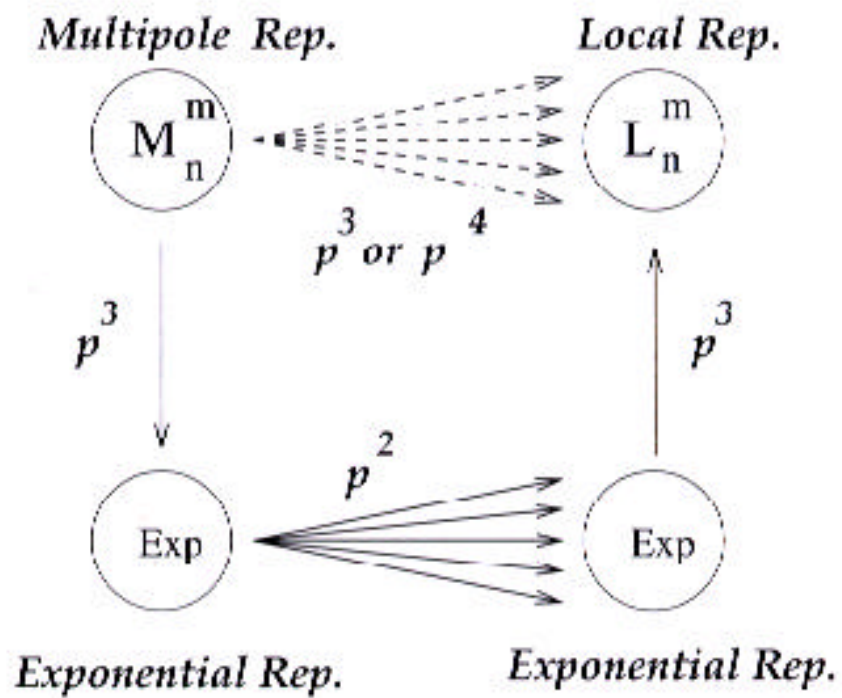
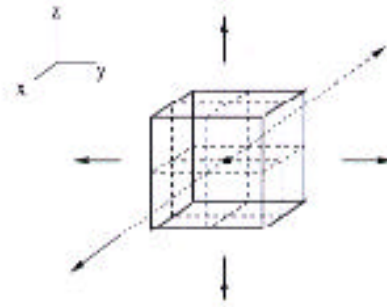
$$\sum_{n=0}^p \sum_{m=-n}^n \frac{M_n^m Y_n^m(\theta, \phi)}{r^{n+1}} \approx \sum_{j=1}^{P_l} \sum_{k=1}^{K_j} e^{-\lambda_j(z - ix \cos \theta_k - iy \theta_k)} S(j, k)$$



## Exponential representation

Precision	p	Exp. Basis Fns.
$10^{-3}$	10	52
$10^{-6}$	19	258
$10^{-9}$	29	670

# Exponential Translation



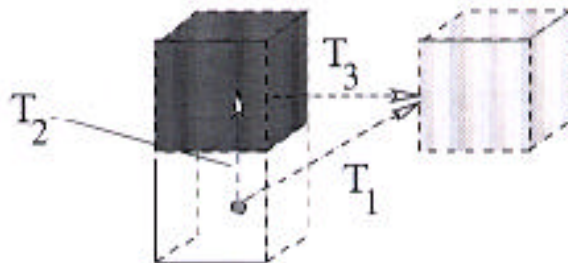
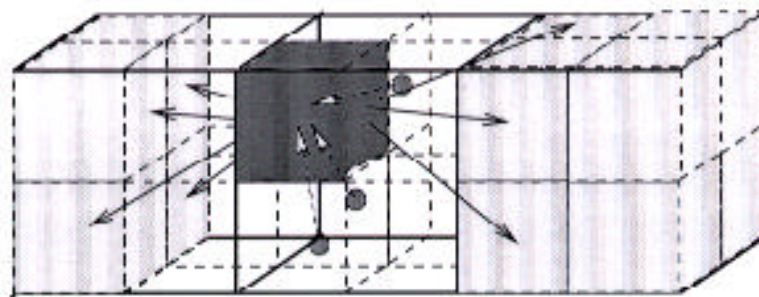
## Operation Count

$$189 \frac{N}{s} p^2 + 2 N p^2 + 54 N s + 6 \frac{N}{s} p^3.$$

Setting  $s = 2p$ , the total operation count is

$$200 N p + 5 N p^2.$$

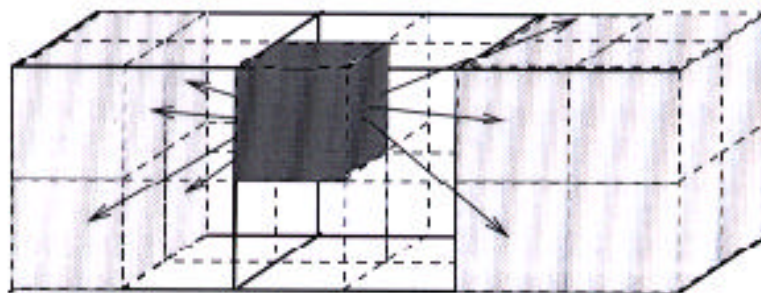
## Reducing the Interaction List



- ▷ Diagonal operators commute
- ▷  $T_1 = T_3 \cdot T_2$
- ▷ Merge before translation
- ▷ Reduces number of interactions per box to  $\leq 40$

## Sweeping Under the Rug:

- ▷ Numerical compression of translation operators
- ▷ Harmonics exterior to a truncated cylinder, harmonics interior to a truncated cylinder, harmonics exterior to union of two truncated cones, etc.
- ▷ Nasty formulae, fairly simple numerical schemes
- ▷ A lot of fuss for a factor of two or so



## Operation Count

$$40 \frac{N}{s} p^2 + 2 N p^2 + 54 N s + 6 \frac{N}{s} p^3.$$

Setting  $s = 1.5 p$ , the total operation count is  
 $\approx$

$$100 N p + 6 N p^2.$$

## Random Distribution Inside a Cube

3-digit accuracy, times in seconds on  
UltraSPARC 1, 167 Mhz; calculations  
performed in single precision

$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20000	4	13.3	233	$7.9 \cdot 10^{-4}$
50000	4	24.7	1483	$5.2 \cdot 10^{-4}$
200000	5	158	24330	$8.4 \cdot 10^{-4}$
500000	5	268	138380	$7.0 \cdot 10^{-4}$
1000000	6	655	563900	$7.1 \cdot 10^{-4}$



## Random Distribution Inside a Cube

6-digit accuracy, calculations performed in single precision

$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20000	3	15.9	233	$5.1 \cdot 10^{-7}$
50000	4	69	1483	$2.8 \cdot 10^{-7}$
200000	4	198	24330	$4.9 \cdot 10^{-7}$
500000	5	586	138380	$4.4 \cdot 10^{-7}$
1000000	5	1245	563900	$4.4 \cdot 10^{-7}$



## Random Distribution Inside a Cube

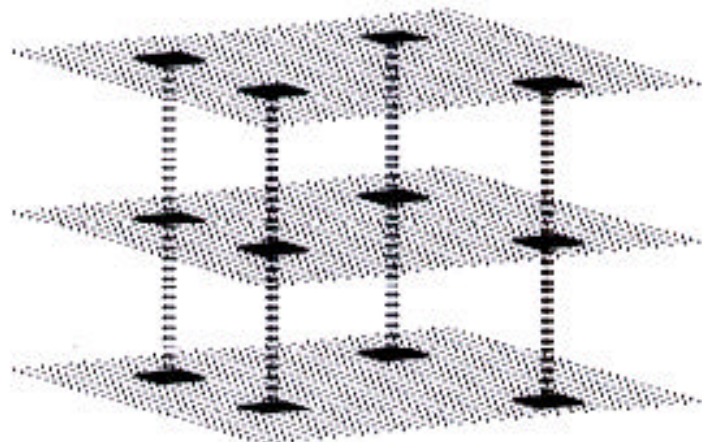
9-digit accuracy, calculations performed in double precision

$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20000	3	34	296	$2.8 \cdot 10^{-10}$
50000	3	96	1920	$1.6 \cdot 10^{-10}$
200000	4	385	30800	$1.6 \cdot 10^{-10}$
500000	4	1219	192600	$1.2 \cdot 10^{-10}$

## Distribution On a Complicated Surface

3-digit accuracy, calculations performed in single precision

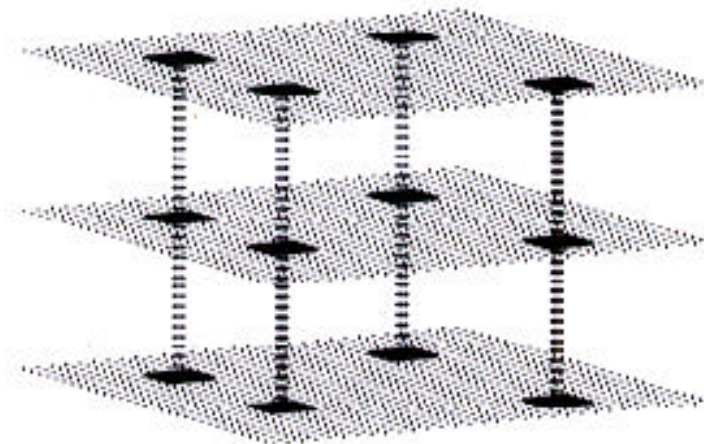
$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20880	7	6.7	243	$2.2 \cdot 10^{-4}$
51900	8	17	1539	$2.7 \cdot 10^{-4}$
203280	9	60	24730	$3.4 \cdot 10^{-4}$
503775	10	164	141060	$3.3 \cdot 10^{-4}$
1007655	10	282	568090	$2.9 \cdot 10^{-4}$



## Distribution On a Complicated Surface

6-digit accuracy, calculations performed in single precision

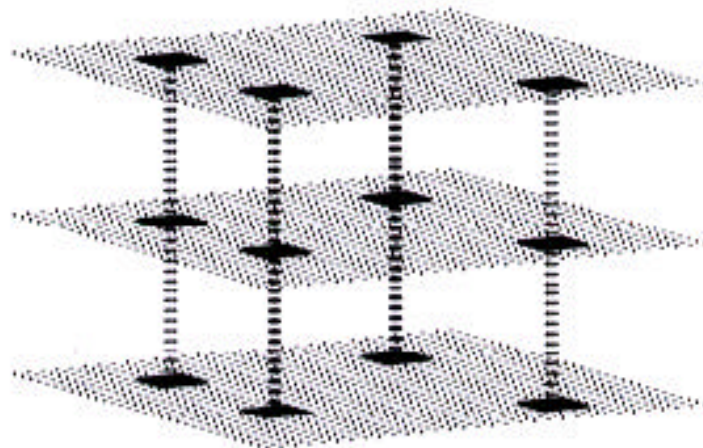
$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20880	7	17	243	$1.3 \cdot 10^{-7}$
51900	8	40	1539	$9.8 \cdot 10^{-8}$
203280	9	149	24730	$1.2 \cdot 10^{-7}$
503775	9	323	141060	$2.6 \cdot 10^{-7}$
1007655	10	714	568090	$2.0 \cdot 10^{-7}$



## Distribution On a Complicated Surface

9-digit accuracy, calculations performed in double precision

$N$	Levels	$T_{fmm}$	$T_{dir}$	Error
20880	6	46	309	$3.6 \cdot 10^{-12}$
51900	7	101	2020	$1.1 \cdot 10^{-10}$
203280	8	342	32050	$6.5 \cdot 10^{-12}$
503775	9	896	193900	$1.0 \cdot 10^{-11}$





## Observations

- ▷ For uniform structures (worst case):
- ▷ Breakeven point less than 1000 for 3 to 4 digit accuracy
- ▷ Breakeven point around 2000 for 6 digit accuracy
- ▷ Breakeven point around 3000 for 9 digit accuracy
- ▷ No loss of accuracy due to adaptivity
- ▷ A black box, as per original plan
- ▷ Large-scale problems manageable on desktop computers

## Post-Mortem

- ▷ A simple formulation (gravitational n-body problem, integral equations of classical potential theory)
- ▷ Fairly simple incantational solution (early FMM schemes)
- ▷ The scheme becomes somewhat involved technically before becoming useful for anything
- ▷ Combination of a little mathematics and a fair amount of engineering
- ▷ Temptation to be a crook
- ▷ We were lucky

## Now What ?

- ▷ A different set of bottlenecks and trade-offs: discretization, convergence, etc.
- ▷ Other potentials: Helmholtz, Yukawa, Hea Wave Equation,...
- ▷ Helmholtz potentials: at low frequencies similar to Laplace; at high frequencies quite different. In all regimes not quite as simple as Laplace
- ▷ Different types of equations: parabolic, hyperbolic, etc.
- ▷ Black boxes all
- ▷ Applications, modifications, etc.
- ▷ There are still some freebies left