

Communication Complexity of the Fast Multipole Method and its Algebraic Variants

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Fast Direct Solvers for Elliptic PDEs
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Computational Complexity $\mathcal{O}(N^2) \rightarrow \mathcal{O}(N)$

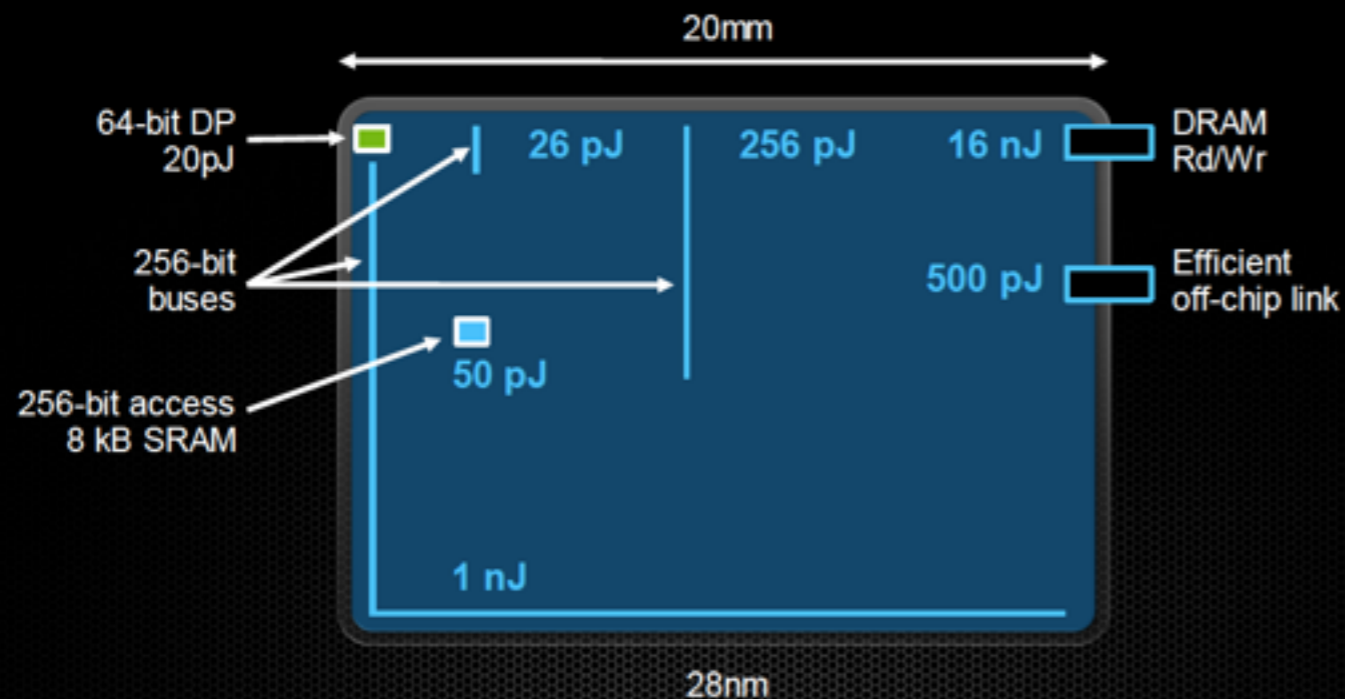
Given a problem of size N , how many arithmetic operations does the algorithm perform?

Communication Complexity $\mathcal{O}(P) \rightarrow \mathcal{O}(\log P)$

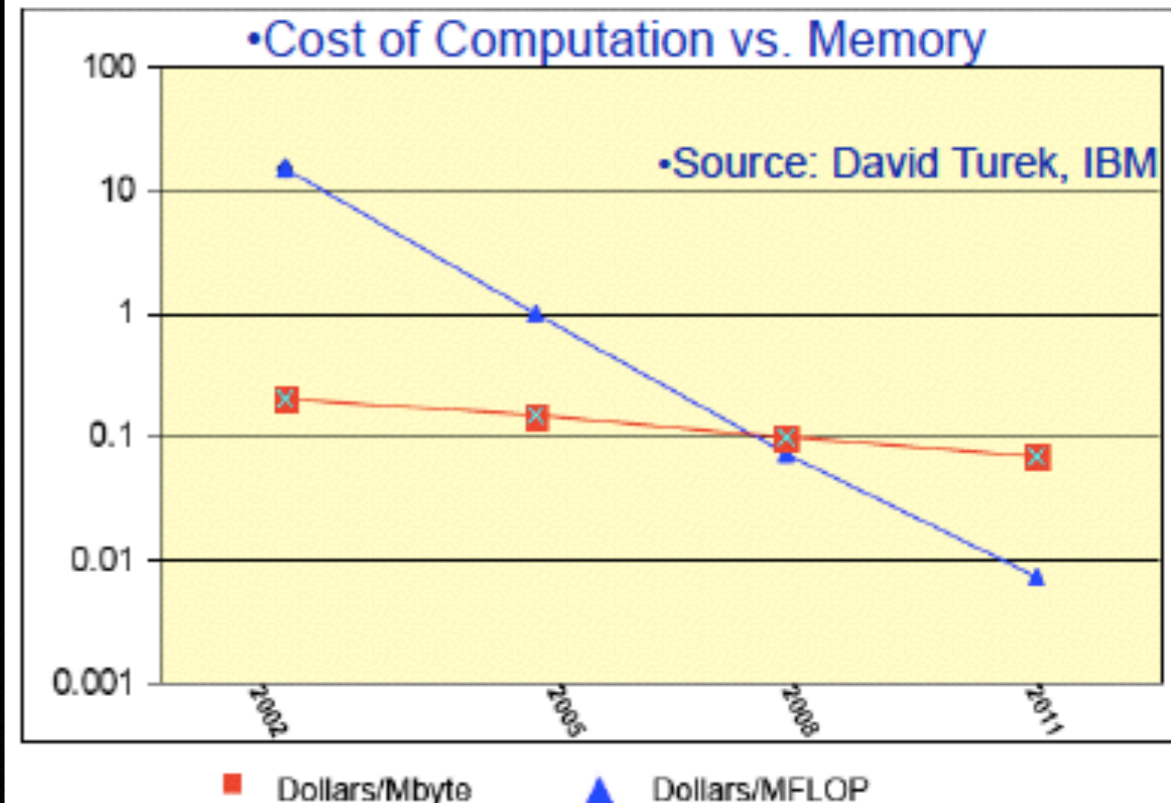
Given a problem on P parallel processes, how much data does the algorithm have to send?

The High Cost of Data Movement

Fetching operands costs more than computing on them



Source: Bill Dally, NVIDIA



Communication Complexity of FMM

SIAM J. SCI. COMPUT.
Vol. 19, No. 2, pp. 635–656, March 1998

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PROVABLY GOOD PARTITIONING AND LOAD BALANCING ALGORITHMS FOR PARALLEL ADAPTIVE N-BODY SIMULATION*

SHANG-HUA TENG†

A massively parallel adaptive fast-multipole method on heterogeneous architectures

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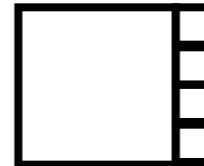
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$$\sum_{i=0}^{\log P-1} 2^i = \mathcal{O}(P)$$

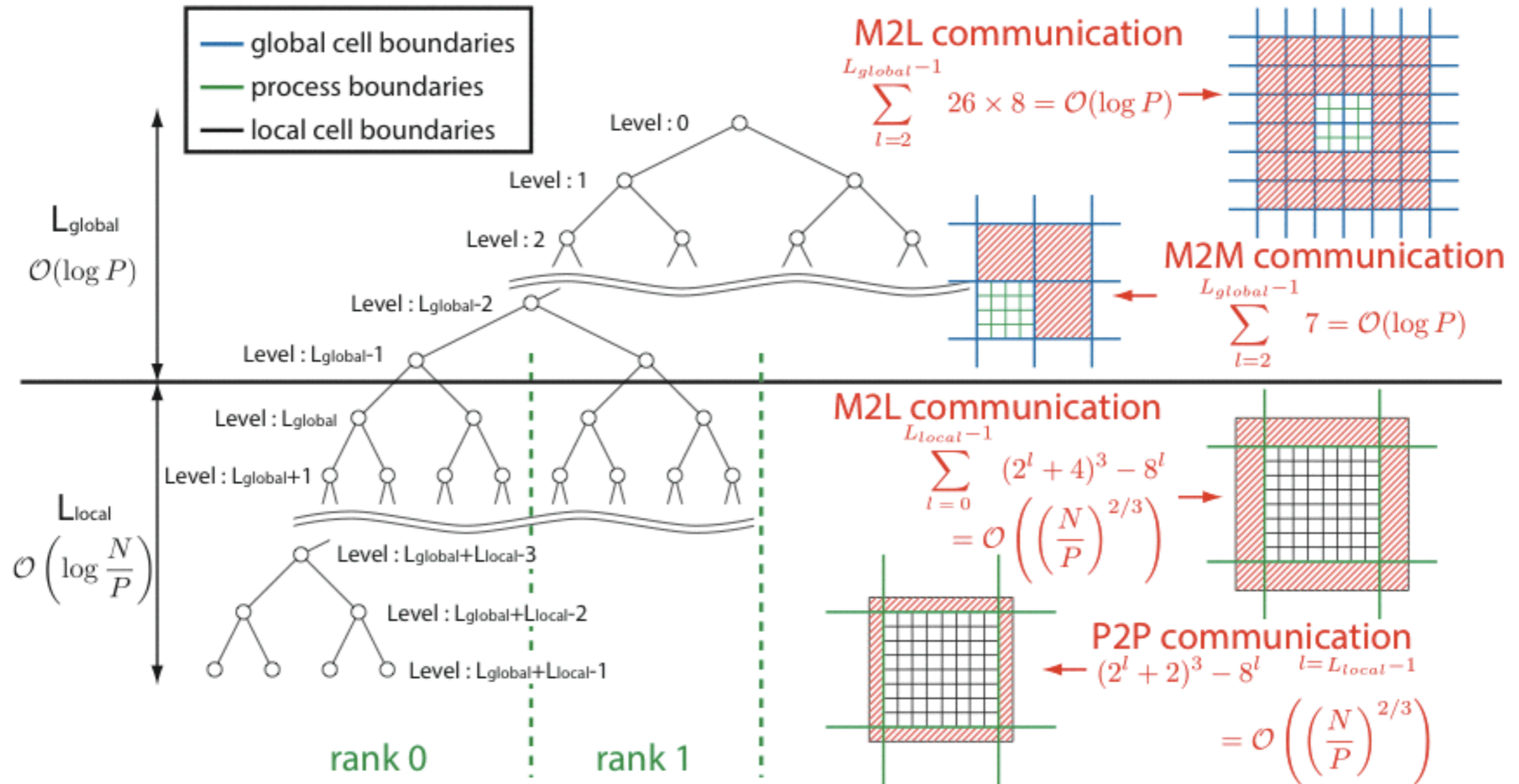


$$\sum_{i=0}^{\log P-1} \min(2^{\log P-i-1}, 2^i) = \mathcal{O}(\sqrt{P})$$



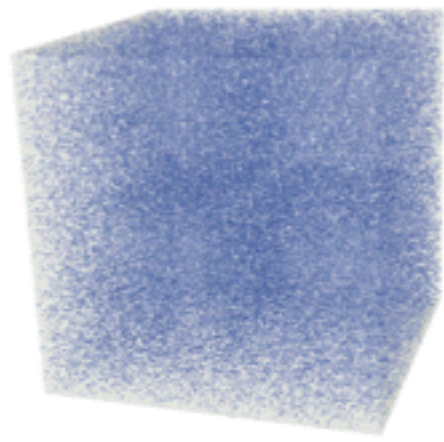
Reference	Processes		Data per Process		Communication complexity
Teng	$\mathcal{O}(P)$		$\mathcal{O}((N/P)^{2/3}(\log N + \mu)^{1/3})$		$\mathcal{O}(P(N/P)^{2/3}(\log N + \mu)^{1/3})$
Lashuk <i>et al.</i>	$\mathcal{O}(\sqrt{P})$		$\mathcal{O}((N/P)^{2/3})$		$\mathcal{O}(\sqrt{P}(N/P)^{2/3})$
Ibeid <i>et al.</i>	Global	Local	Global	Local	Global + Local
	$\mathcal{O}(\log P)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}((N/P)^{2/3})$	$\mathcal{O}(\log P + (N/P)^{2/3})$

Uniform Case

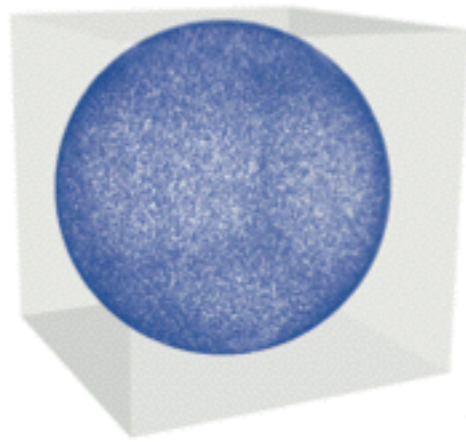


	Processes	Cells per level	Cells per Process	Communication
Global M2L	$\sum_i^{\log P} 26$	26×8	8	$\mathcal{O}(\log P)$
Global M2M	$\sum_i^{\log P} 7$	7	1	$\mathcal{O}(\log P)$
Local M2L	26	$(2^i + 4)^3 - 8^i$	$\sum_i^{\log_s(N/P)} (2^i + 4)^3 - 8^i$	$\mathcal{O}((N/P)^{2/3})$
Local P2P	26	$(2^i + 2)^3 - 8^i$	$(2^{\log_s(N/P)} + 2)^3 - 8^{\log_s(N/P)}$	$\mathcal{O}((N/P)^{2/3})$

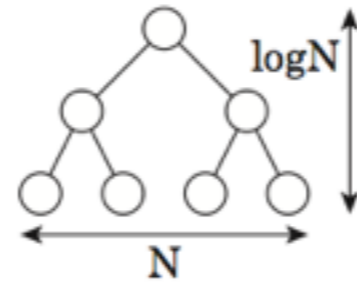
Nonuniform Case



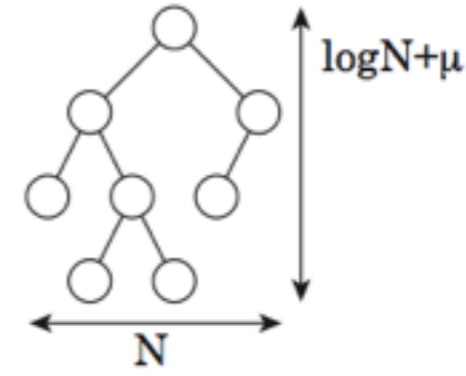
(a) Cube



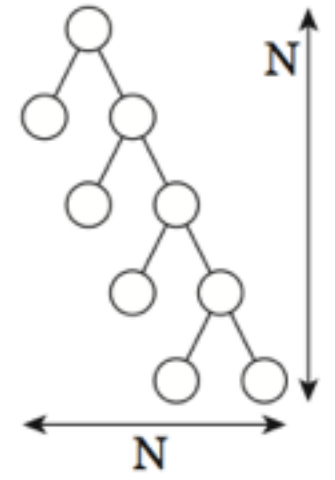
(b) Sphere



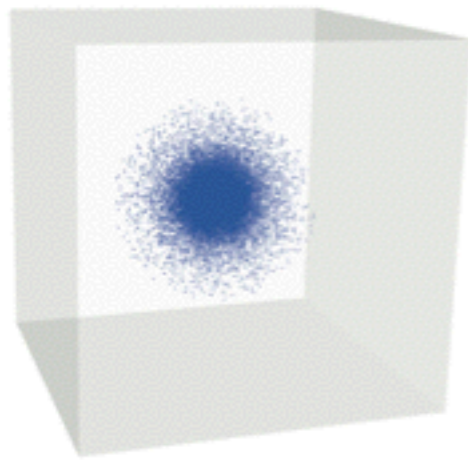
(a) Uniform



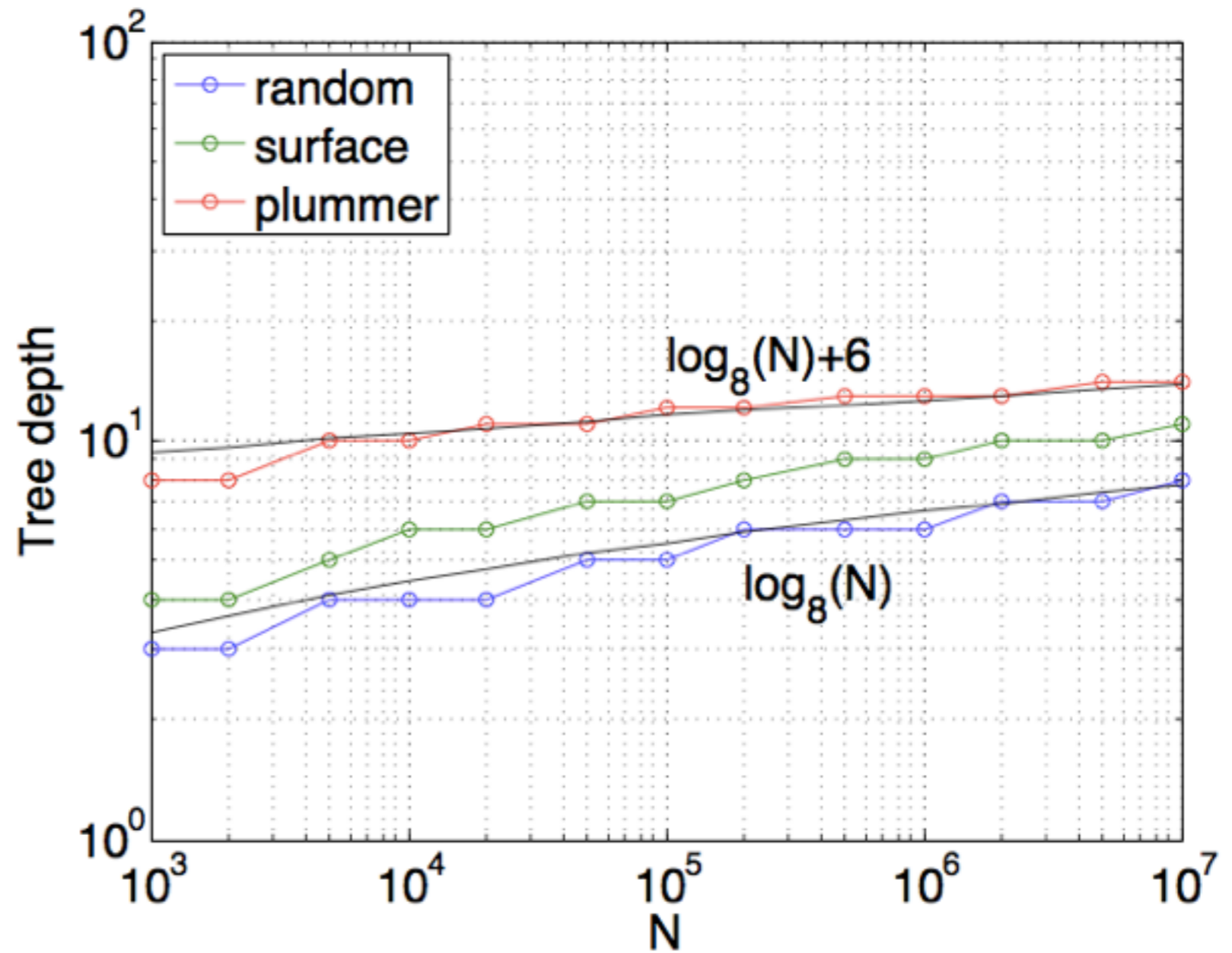
(b) Nonuniform



(c) Pathological



(c) Plummer



Nonuniform Case

	Processes	Cells per level	Cells per Process	Communication
Global M2L	$\sum_i^{\log P} 26$	26×8	8	$\mathcal{O}(\log P)$
Global M2M	$\sum_i^{\log P} 7$	7	1	$\mathcal{O}(\log P)$
Local M2L	26	$(2^i + 4)^3 - 8^i$	$\sum_i^{\log_s(N/P)} (2^i + 4)^3 - 8^i$	$\mathcal{O}((N/P)^{2/3})$
Local P2P	26	$(2^i + 2)^3 - 8^i$	$(2^{\log_s(N/P)} + 2)^3 - 8^{\log_s(N/P)}$	$\mathcal{O}((N/P)^{2/3})$



	Processes	Cells per level	Cells per Process	Communication
Global M2L	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Global M2M	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Local M2L	$\mathcal{O}(1)$	$\mathcal{O}(4^i)$	$\sum_i^{\log_s(N/P)} \mathcal{O}(4^i)$	$\mathcal{O}((N/P)^{2/3})$
Local P2P	$\mathcal{O}(1)$	$\mathcal{O}(4^i)$	$\mathcal{O}(4^{\log_s(N/P)})$	$\mathcal{O}((N/P)^{2/3})$

Algebraic Case (Mat-Vec)

$$y = \left(\sum_{(i,j) \in D} A_{ij} \right) x + \left(\sum_{(i,j) \in L} U_i S_{ij} V_j^t \right) x = \underbrace{\sum_{(i,j) \in D} A_{ij} x_j}_{\text{Dense mat-vecs operations}} + \underbrace{\sum_{i \in I} U_i \sum_{(i,j) \in L} S_{ij} \underbrace{V_j^t x}_{\text{Upsweep}}}_{\text{Coupling phase}}$$

Downsweep

	Processes	Cells per level	Cells per Process	Communication
Global M2L	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Global M2M	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Local M2L	$\mathcal{O}(1)$	$\mathcal{O}(4^i)$	$\sum_i^{\log_s(N/P)} \mathcal{O}(4^i)$	$\mathcal{O}((N/P)^{2/3})$
Local P2P	$\mathcal{O}(1)$	$\mathcal{O}(4^i)$	$\mathcal{O}(4^{\log_s(N/P)})$	$\mathcal{O}((N/P)^{2/3})$



Matrix Operation	FMM operation	Processes	Blocks per level	Blocks per Process	Communication
Global $\sum S_{ij} \hat{x}_j$	Global M2L	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Global $\sum F_k^t \hat{x}_k$	Global M2M	$\sum_i^{\log P} \mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}(\log P)$
Local $\sum S_{ij} \hat{x}_j$	Local M2L	$\mathcal{O}(1)$	$\mathcal{O}(2^{(d-1)i})$	$\sum_i^{\log_{2^d}(N/P)} \mathcal{O}(2^{(d-1)i})$	$\mathcal{O}((N/P)^{\frac{d-1}{d}})$
Local $\sum A_{ij} x_j$	Local P2P	$\mathcal{O}(1)$	$\mathcal{O}(2^{(d-1)i})$	$\mathcal{O}(2^{(d-1) \log_{2^d}(N/P)})$	$\mathcal{O}((N/P)^{\frac{d-1}{d}})$

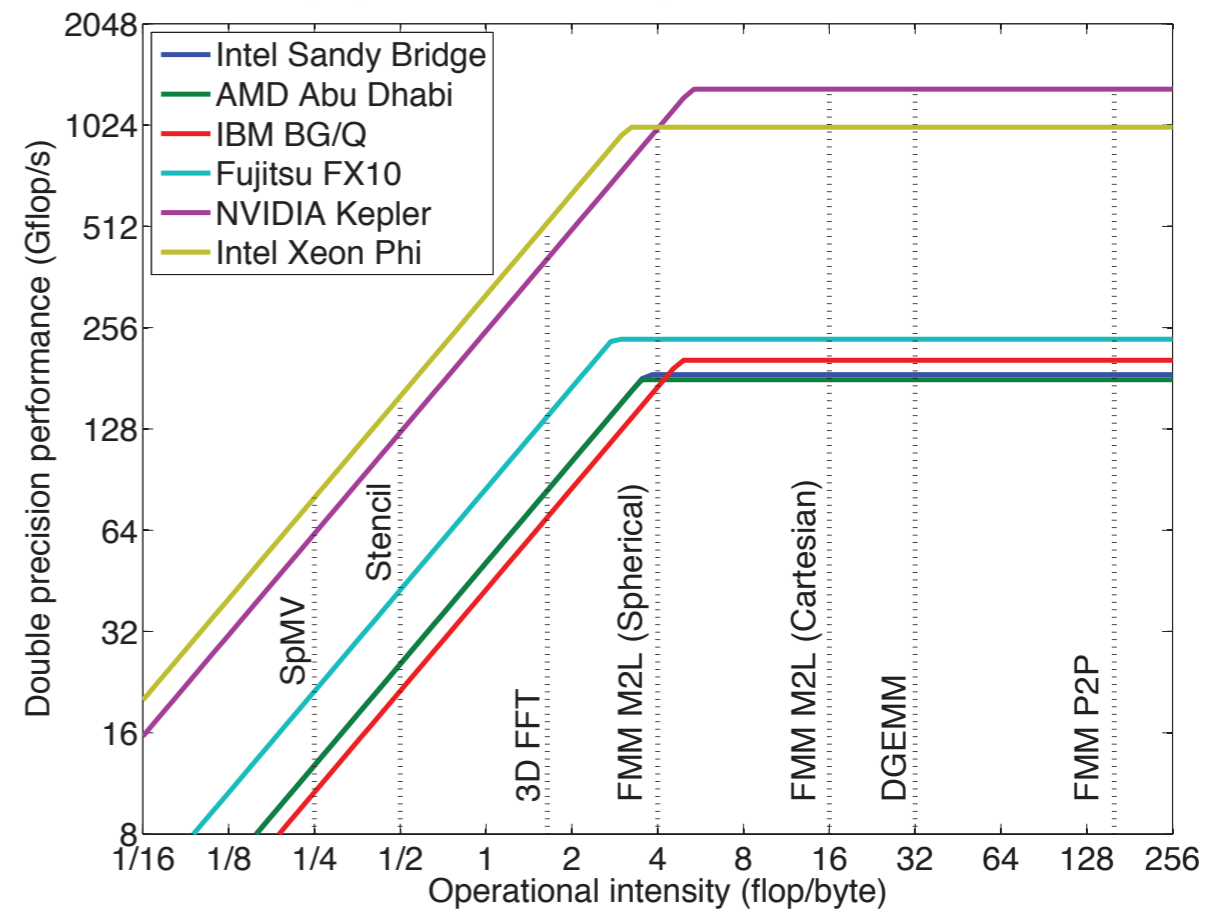
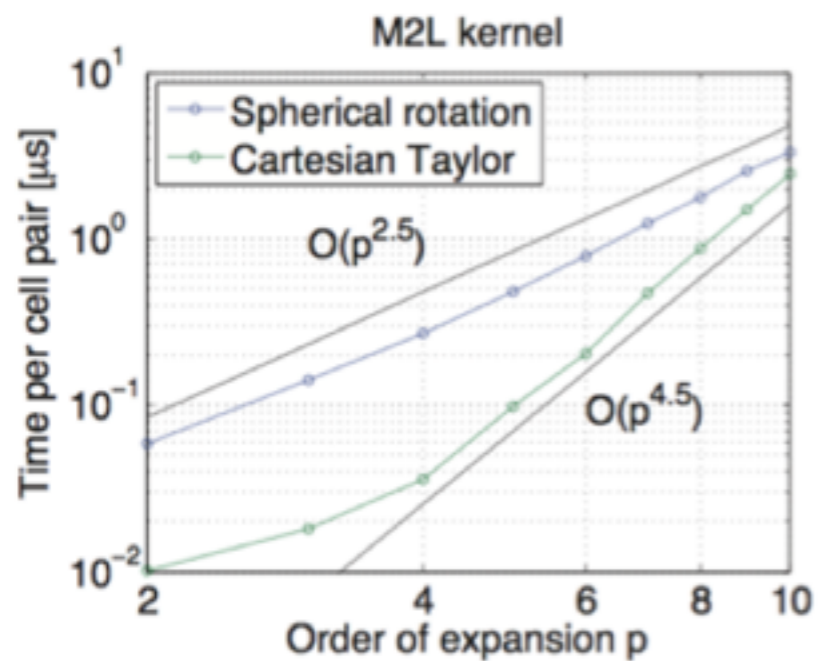
Asymptotic Constants

FMM

Type of expansion (+M2L acceleration)	Storage	Arithmetic
Cartesian Taylor	$\mathcal{O}(p^3)$	$\mathcal{O}(p^6)$
Cartesian Chebychev	$\mathcal{O}(p^3)$	$\mathcal{O}(p^6)$
Spherical harmonics	$\mathcal{O}(p^2)$	$\mathcal{O}(p^4)$
Spherical harmonics+rotation	$\mathcal{O}(p^2)$	$\mathcal{O}(p^3)$
Spherical harmonics+FFT	$\mathcal{O}(p^2)$	$\mathcal{O}(p^2 \log^2 p)$
Planewave	$\mathcal{O}(p^2)$	$\mathcal{O}(p^3)$
Equivalent charges	$\mathcal{O}(p^2)$	$\mathcal{O}(p^4)$
Equivalent charges+FFT	$\mathcal{O}(p^3)$	$\mathcal{O}(p^3 \log p)$

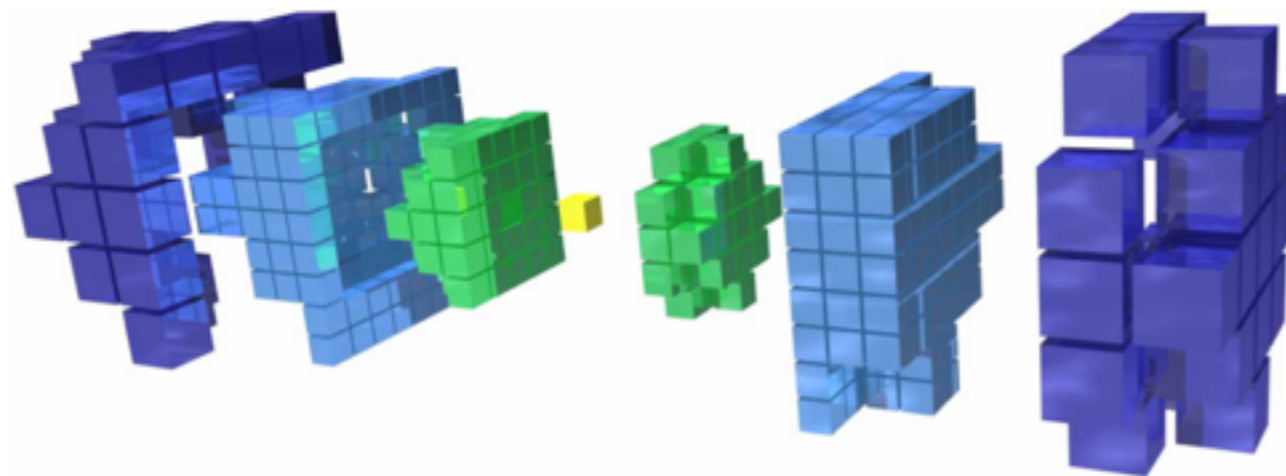
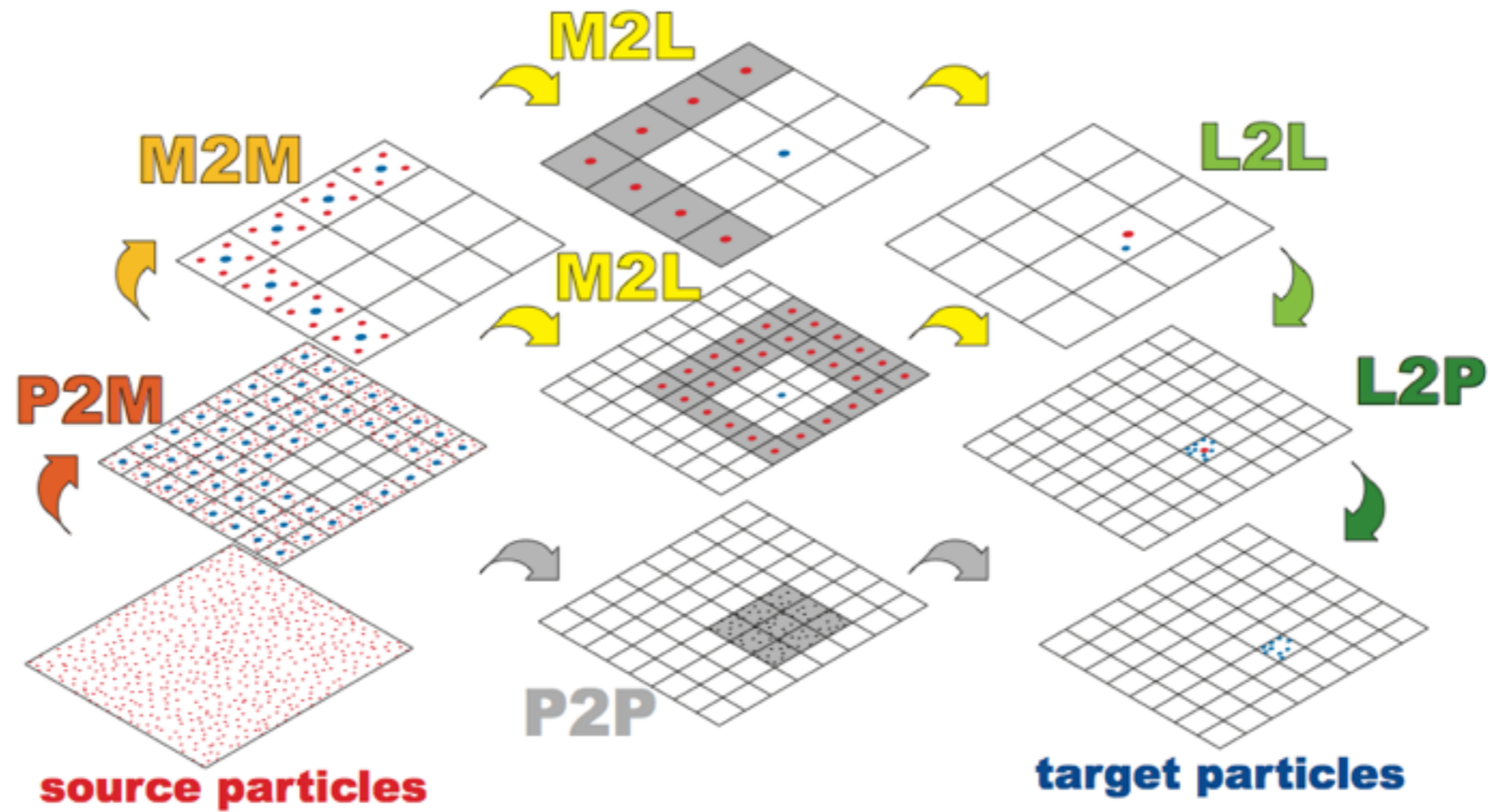
H-matrix

Type of low rank approximation	Reference
Rank-revealing LU	Pan (2000)
Rank-revealing QR	Gu & Eisenstat (1996)
Pivoted QR	Kong et al. (2011)
Truncated SVD	Grasedyck & Hackbusch (2003)
Randomized SVD	Liberty et al. (2007)
Adaptive cross approximation	Rjasanow (2002)
Hybrid cross approximation	Börm (2005)
Chebychev interpolation	Dutt et al. (1996)



Better complexity \neq Better performance

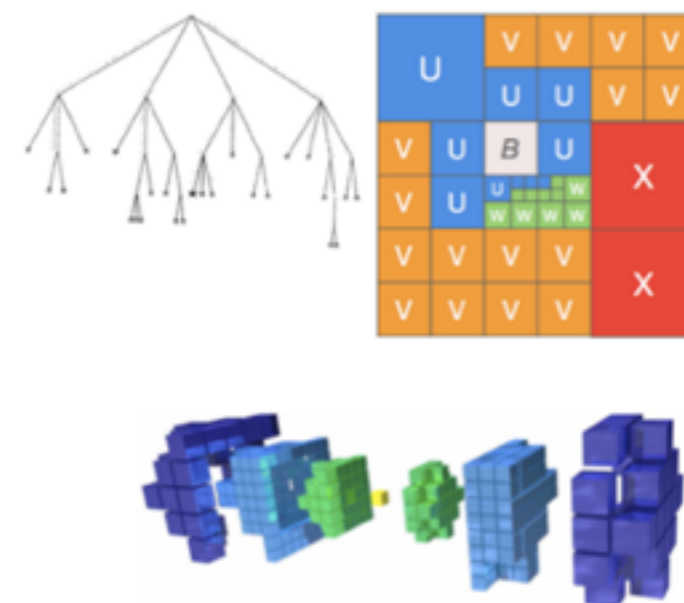
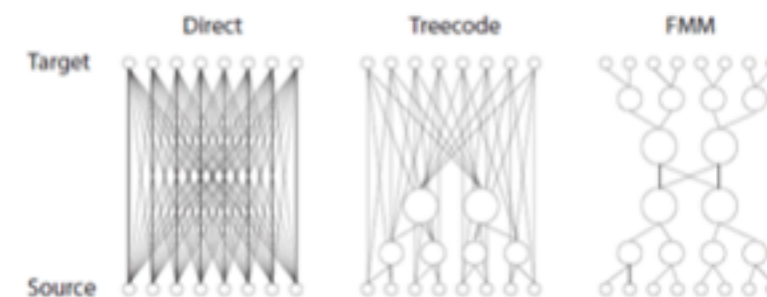
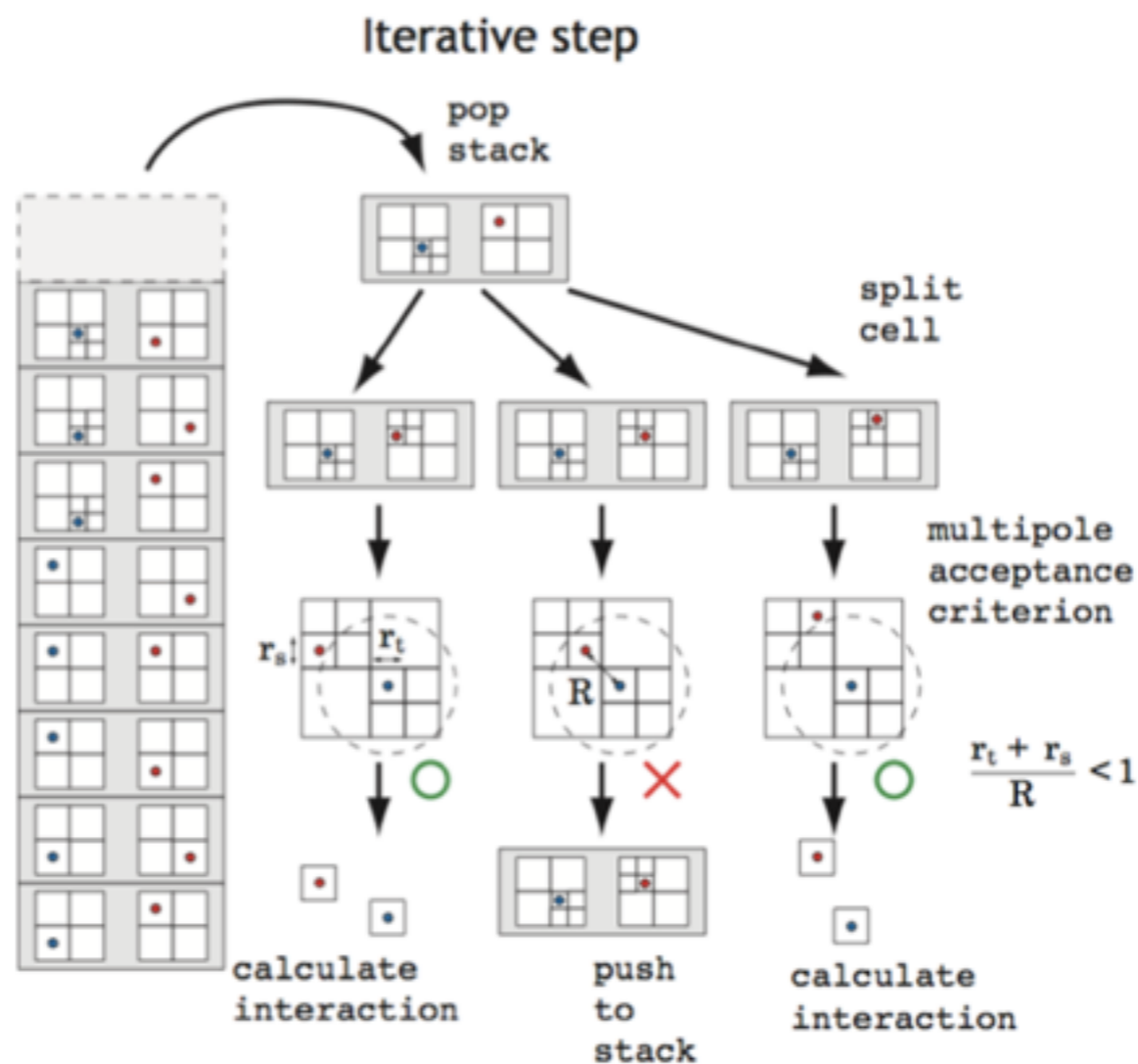
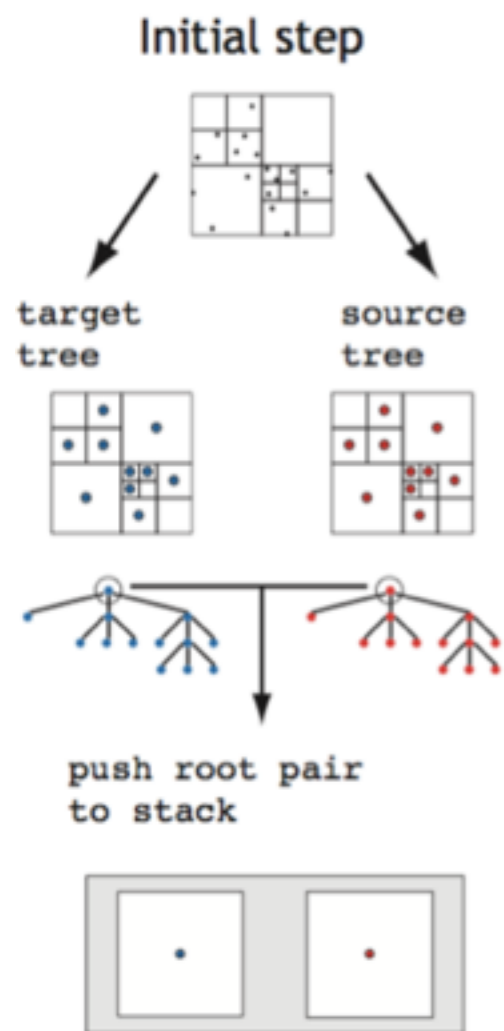
Finding Admissible Blocks



Neighbor List Per Level Per Block?

Dual Tree Traversal

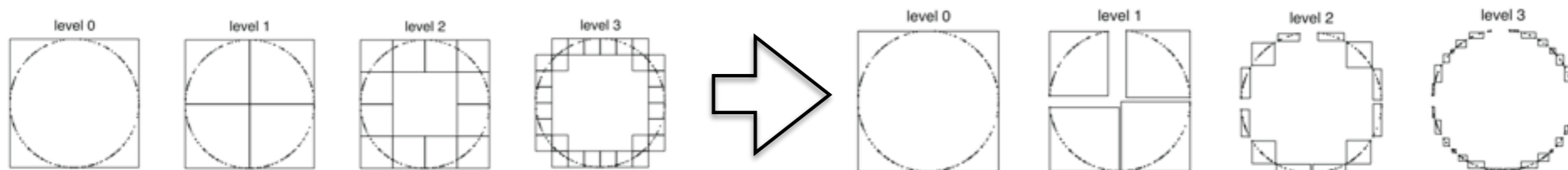
W. Dehnen, J. Comput. Phys. 179; 27-42 (2002)



No Neighbor List Needed

S.-H. Teng, SIAM J. Sci. Comput. 19; 635-656 (1998)

Cells don't have to be cubic



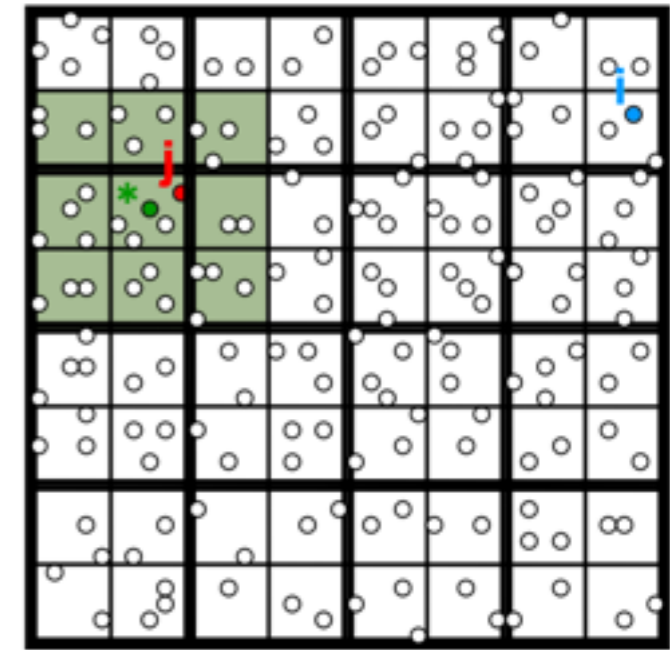
Dual Tree Traversal


- No need to explicitly form interaction lists
- The definition of well-separatedness (size of neighbor region) can be adjusted flexibly without modifying the code
- The resulting neighbor region and M2L interaction region naturally have a spherical shape (as opposed to the suboptimal cubic shape)
- It is applicable to adaptive trees without any modification
- It lends itself to MPI parallelization without any modification (by simply using the local essential tree as the source tree)
- It is easily extendable to periodic boundary conditions (by traversing all periodic image trees as the source tree)
- It can handle mutual M2L interaction, and can satisfy Newton's third law (M2L is neither target centric nor source centric, but completely symmetric)
- It works well with task based threading tools like Intel TBB, Cilk, MassiveThreads, etc., where tasks are spawned while the tree is traversed
- The cells don't have to be cubic. For example, high aspect ratio rectangles or hierarchical K-means is permitted.
- It can be implemented in less than 100 lines of code, and is therefore trivial to debug

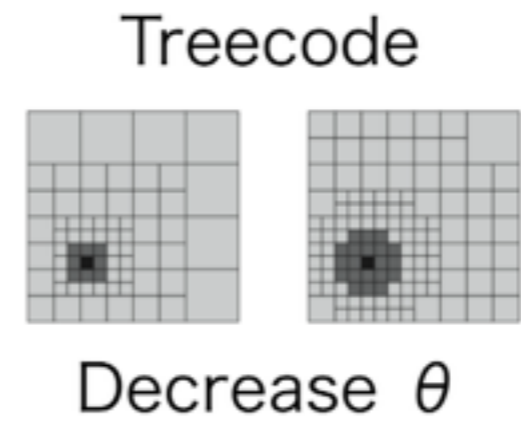
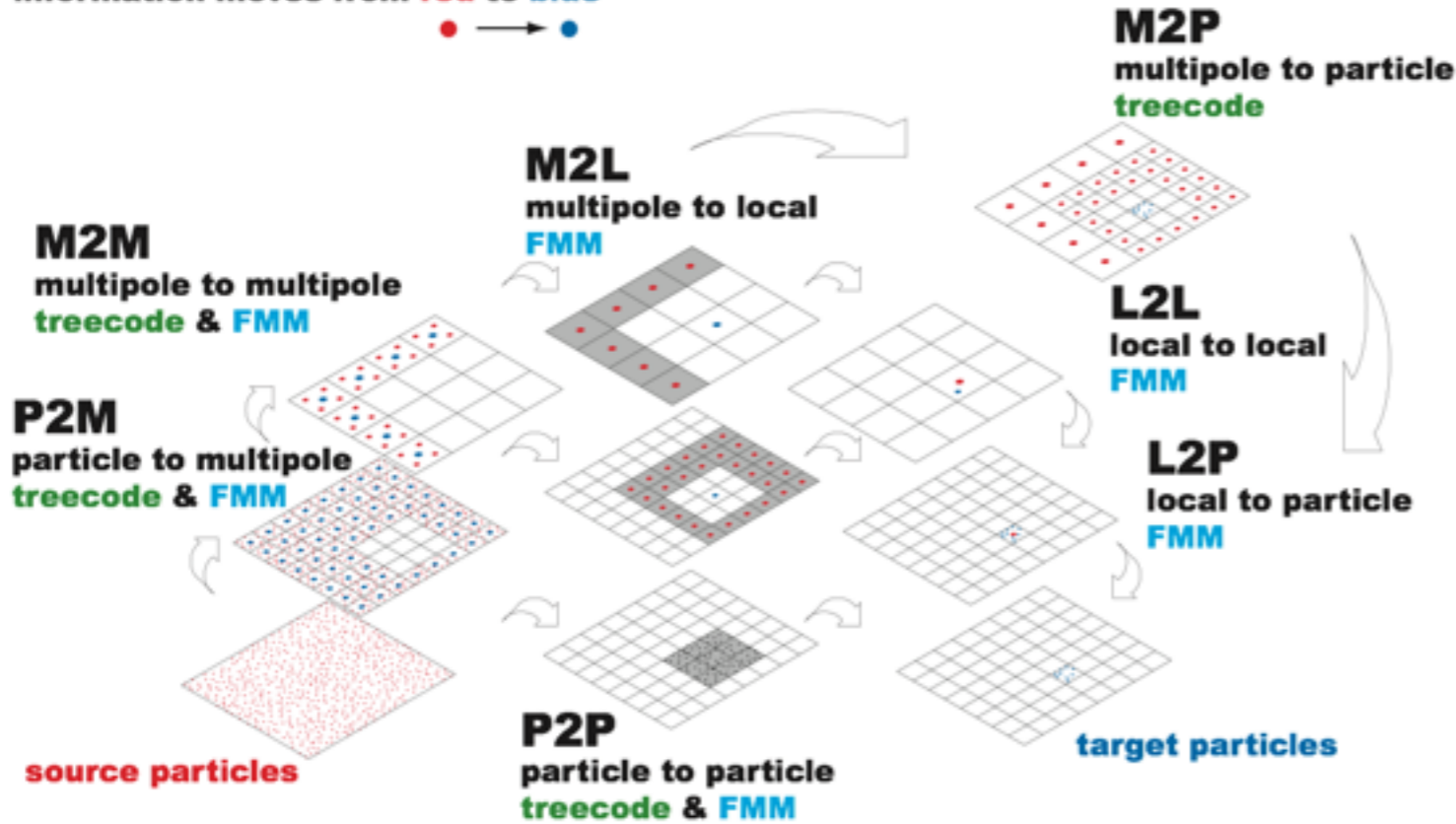
h-p FMM

$$\theta = \frac{|\mathbf{x}_j - \mathbf{x}_*|}{|\mathbf{x}_i - \mathbf{x}_*|} < 1$$

Error bound : $O(\theta^p)$



information moves from red to blue




FMM

$$\frac{1}{1-\theta} = \sum_{k=0}^{p-1} \theta^k$$

Increase p

	$p = 3$	$p = 4$	$p = 5$	$p = 6$
$Err = 10^{-2}$	$\theta = 1.00$ $time = 0.016s$	$\theta = 1.18$ $time = 0.012s$	$\theta = 1.23$ $time = 0.015s$	$\theta = 1.24$ $time = 0.026s$
$Err = 10^{-3}$	$\theta = 0.67$ $time = 0.036s$	$\theta = 0.78$ $time = 0.027s$	$\theta = 0.91$ $time = 0.024s$	$\theta = 0.94$ $time = 0.038s$
$Err = 10^{-4}$	$\theta = 0.30$ $time = 0.22s$	$\theta = 0.49$ $time = 0.085s$	$\theta = 0.62$ $time = 0.071s$	$\theta = 0.70$ $time = 0.073s$
$Err = 10^{-5}$	$\theta = 0.12$ $time = 1.38s$	$\theta = 0.20$ $time = 0.59s$	$\theta = 0.36$ $time = 0.21s$	$\theta = 0.45$ $time = 0.21s$

Spatially Varying “h” or “p”

Corrected Article: “An error-controlled fast multipole method”
[J. Chem. Phys. 131, 244102 (2009)]

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Spatially varying rank

Journal of Computational Physics **179**, 27–42 (2002)
doi:10.1006/jcph.2002.7026

A Hierarchical $\mathcal{O}(N)$ Force Calculation Algorithm

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Spatially varying admissibility



SIMD friendly

Conclusions

Reference	Processes		Data per Process		Communication complexity
Teng	$\mathcal{O}(P)$		$\mathcal{O}((N/P)^{2/3}(\log N + \mu)^{1/3})$		$\mathcal{O}(P(N/P)^{2/3}(\log N + \mu)^{1/3})$
Lashuk <i>et al.</i>	$\mathcal{O}(\sqrt{P})$		$\mathcal{O}((N/P)^{2/3})$		$\mathcal{O}(\sqrt{P}(N/P)^{2/3})$
Ibeid <i>et al.</i>	Global	Local	Global	Local	Global + Local
	$\mathcal{O}(\log P)$	$\mathcal{O}(1)$	$\mathcal{O}(1)$	$\mathcal{O}((N/P)^{2/3})$	$\mathcal{O}(\log P + (N/P)^{2/3})$

- We have proved a new upper bound for the communication complexity of FMM
- The dual tree traversal allows hierarchical methods to use adaptive admissibility conditions, which results in fine grain uniformity that is easier to vectorize

Title