

A new adaptive node-cluster algorithm in Fast Multipole Method

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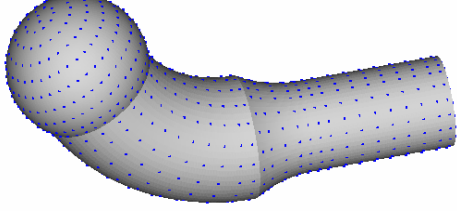
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Hybrid BNM

➤ Main features:

- Combines a modified functional with the *Moving Least Squares* (MLS) approximation
- Three independent variables
 - internal temperature
 - boundary temperature
 - boundary normal flux



Example of meshless discretization

➤ Variables approximation

- Domain variables
- Boundary variables

$$\phi = \sum_{I=1}^N \phi_I^s x_I$$

$$\phi_I^s = \frac{1}{\kappa} \frac{1}{4\pi r(Q, \mathbf{s}_I)}$$

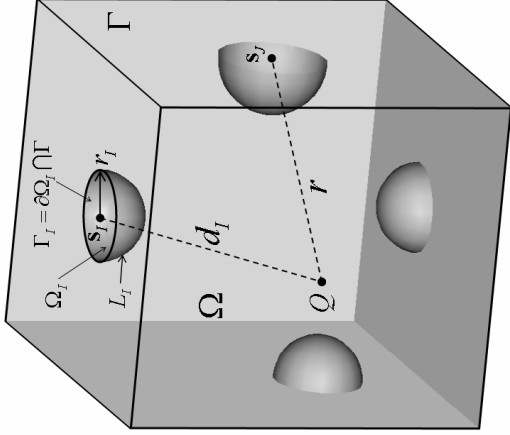
$$\tilde{\phi}(\mathbf{s}) = \sum_{I=1}^N \Phi_I(\mathbf{s}) \hat{\phi}_I$$

$$\tilde{q}(\mathbf{s}) = \sum_{I=1}^N \Phi_I(\mathbf{s}) \hat{q}_I$$



Hybrid BNM (2)

System of equations



$$\mathbf{U}\mathbf{x} = \mathbf{H}\hat{\mathbf{q}}$$

$$U_{IJ} = \int_{\Gamma_s^J} \phi_I^s v_J(Q) d\Gamma$$

$$V_{IJ} = \int_{\Gamma_s^J} q_I^s v_J(Q) d\Gamma$$

$$\mathbf{V}\mathbf{x} = \mathbf{H}\hat{\phi}$$

$$H_{IJ} = \int_{\Gamma_s^J} \Phi_I(\mathbf{s}) v_J(Q) d\Gamma$$

- Three purposes of elements in BEM:
 - To interpolate Boundary variables
 - To facilitate boundary integration
 - To approximate the geometry



Iterative solver

➤ Matrix-vector multiplication

$$x_I'^{k+1} = \sum_{J=1}^N \int_{\Gamma_I} \phi_J^s v_I(Q) x_J^k d\Gamma$$

$$x_I'^{k+1} = \sum_{J=1}^N \int_{\Gamma_I} \frac{\partial \phi_J^s}{\partial n} v_I(Q) x_J^k d\Gamma$$

Complexity of an iterative solver:

Memory: $O(N^2)$ CPU time: $O(N^2)$

Fast Multipole Method (FMM):

Reducing computational complexity to $O(N)$

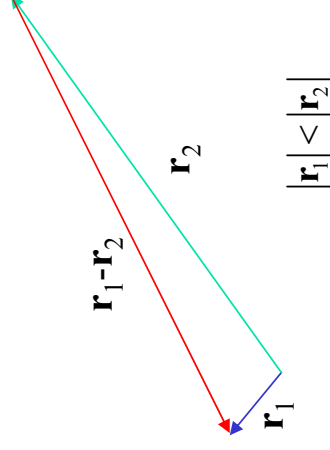


Addition theorems

➤ First addition theorem

Let \mathbf{r}_1 and \mathbf{r}_2 be two vectors with spherical coordinates (r_1, α_1, β_1) and (r_2, α_2, β_2) , respectively. It follows

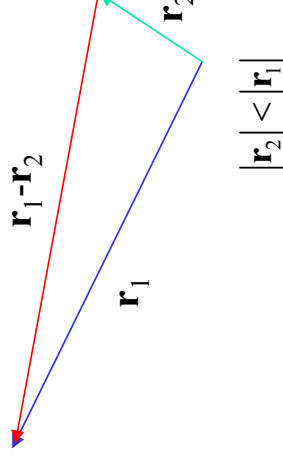
$$\frac{1}{|\mathbf{r}_1 - \mathbf{r}_2|} = \begin{cases} \sum_{n=0}^{\infty} \sum_{m=-n}^n R_n^m(\mathbf{r}_1) \overline{S_n^m(\mathbf{r}_2)}, & |\mathbf{r}_1| < |\mathbf{r}_2| \\ \sum_{n=0}^{\infty} \sum_{m=-n}^n R_n^m(\mathbf{r}_2) \overline{S_n^m(\mathbf{r}_1)}, & |\mathbf{r}_1| > |\mathbf{r}_2| \end{cases}$$



where

$$R_n^m(\mathbf{r}) = \frac{1}{(n+m)!} P_n^m(\cos \alpha) e^{im\beta} r^n$$

$$S_n^m(\mathbf{r}) = (n-m)! P_n^m(\cos \alpha) e^{im\beta} \frac{1}{r^{n+1}}$$



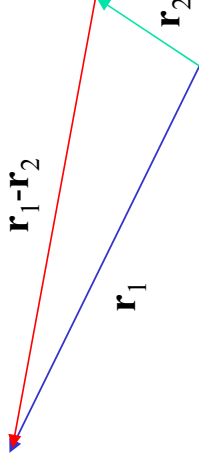


Addition theorems (2)

➤ Second addition theorem

Let \mathbf{r}_1 and \mathbf{r}_2 be two vectors such that $|\mathbf{r}_1| > |\mathbf{r}_2|$, then

$$S_n^m(\mathbf{r}_1 - \mathbf{r}_2) = \sum_{n'=0}^{\infty} \sum_{m'=-n'}^{n'} \overline{R_{n'}^{m'}(\mathbf{r}_2)} S_{n+n'}^{m+m'}(\mathbf{r}_1)$$

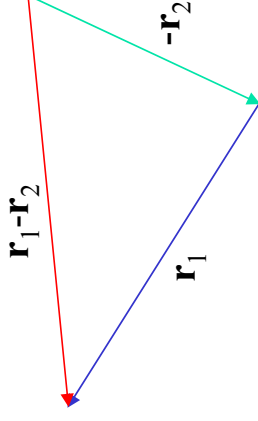


$$|\mathbf{r}_1| > |\mathbf{r}_2|$$

➤ Third addition theorem

Let \mathbf{r}_1 and \mathbf{r}_2 be two arbitrary vectors, then

$$R_n^m(\mathbf{r}_1 - \mathbf{r}_2) = \sum_{n'=0}^n \sum_{m'=-n'}^{n'} R_{n'}^{m'}(-\mathbf{r}_2) R_{n-n'}^{m-m'}(\mathbf{r}_1)$$

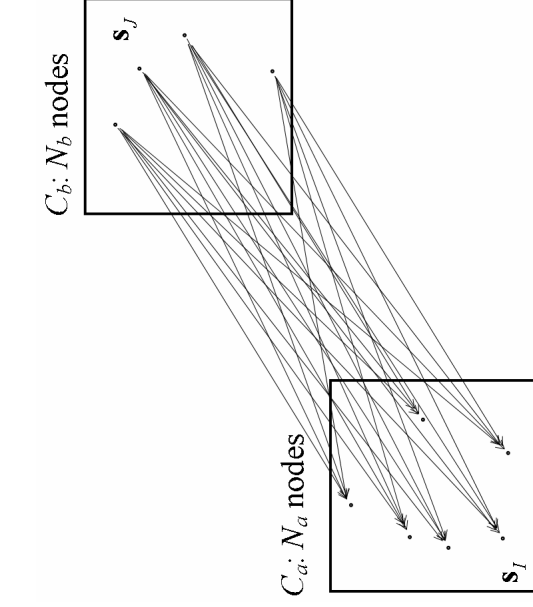




Fast multipole

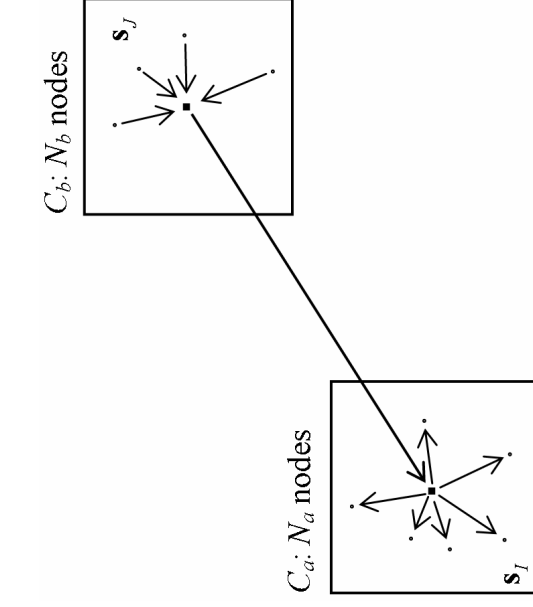
➤ Ideas of FMM

Node-node interactions



Complexity $O(N_a N_b)$

Cell-cell interactions



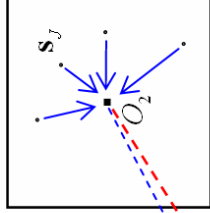
Complexity $O(N_a + N_b)$



Fast multipole (2)

➤ Multipole expansion

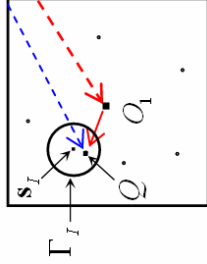
C_b : N_b nodes



$$\phi_J^s = \frac{1}{4\pi K} \frac{1}{r(Q, \mathbf{s}_J)} = \frac{1}{4\pi K} \sum_{n=0}^{\infty} \sum_{m=-n}^n \overline{S_n^m(O_2 Q)} \overline{R_n^m(O_2 \mathbf{s}_J)}$$

for $|\overline{O_2 Q}| > |\overline{O_2 \mathbf{s}_J}|$

C_a : N_a nodes



$$\sum_{J=1}^{N_b} \int_{\Gamma_I} \phi_J^s v_I(Q) x'_J d\Gamma = \sum_{n=0}^{\infty} \sum_{m=-n}^n \int_{\Gamma_I} \frac{1}{4\pi K} \overline{S_n^m(O_2 Q)} v_I(Q) d\Gamma \overline{M_n^m(O_2)}$$

where

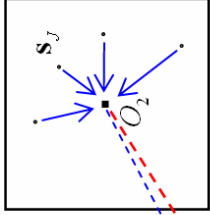
$$\overline{M_n^m(O_2)} = \sum_{J=1}^{N_b} \overline{R_n^m(O_2 \mathbf{s}_J)} x'_J$$



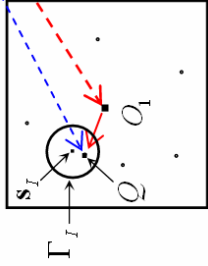
Fast multipole (3)

➤ Local expansion

$C_b: N_b$ nodes



$C_a: N_a$ nodes



$$\overline{S_n^m(O_2 Q)} = \sum_{n'=0}^{\infty} \sum_{m'=-n'}^{n'} (-1)^{n'} R_{n'}^{m'}(\overline{O_1 Q}) \overline{S_{n+n'}^{m+m'}(O_1 O_2)}$$

for $|\overline{O_1 O_2}| > |\overline{O_1 Q}|$

$$\sum_{J=1}^{N_b} \int_{\Gamma_I} \phi_J^s v_I(Q) x'_j d\Gamma = \sum_{n'=0}^{\infty} \sum_{m'=-n'}^{n'} \int_{\Gamma_I} \frac{1}{4\pi K} R_{n'}^{m'}(\overline{O_1 Q}) v_I(Q) d\Gamma L_n^{m'}(O_1)$$

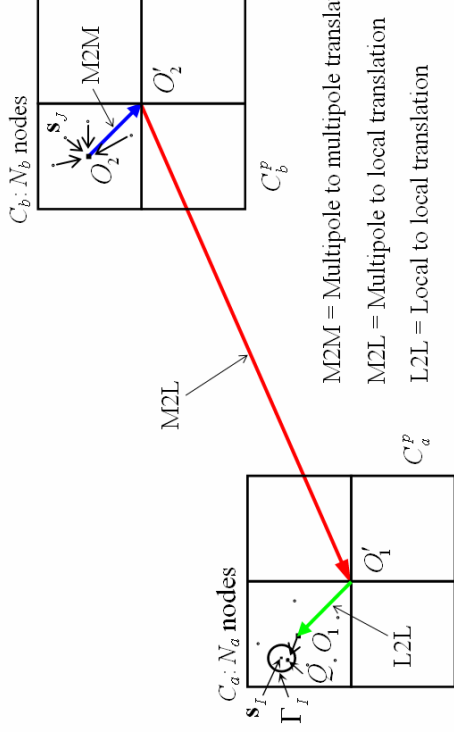
where

$$L_n^{m'}(O_1) = \sum_{n=0}^{\infty} \sum_{m=-n}^n (-1)^{n'} \overline{S_{n+n'}^{m+m'}(O_1 O_2)} M_n^m(Q_2)$$



Fast multipole (4)

► Translation operators



$$M_n^{m'}(Q_2) = \sum_{n=0}^{\infty} \sum_{m=-n}^n R_n^m(\overline{O_2 O_2'}) M_{n-n'}^{m-m'}(Q_2)$$

Multipole to multipole translation

$$L_n^m(O_1') = \sum_{n'=0}^{\infty} \sum_{m'=-n'}^{n'} (-1)^n \overline{S_{n+n'}^{m-m'}}(\overline{O_1' O_2'}) M_{n'}^{m'}(Q_2')$$

Multipole to local translation

$$L_{n'}^{m'}(O_1) = \sum_{n=0}^{\infty} \sum_{m=-n}^n R_{n-n'}^{m-m'}(\overline{O_1 O_1'}) L_n^m(Q_1)$$

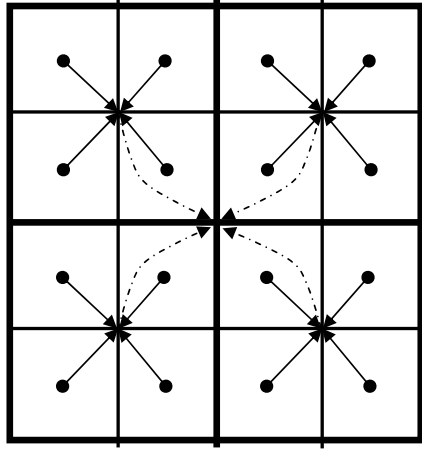
Local to local translation



Fast multipole (5)

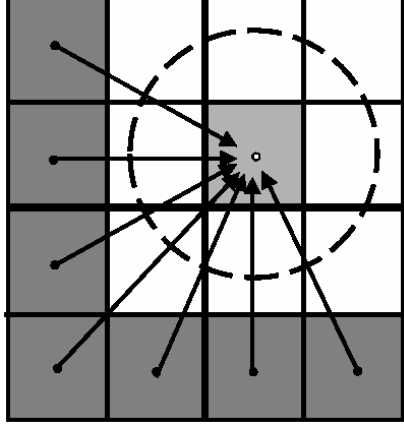
➤ Recursive algorithm

Upward pass

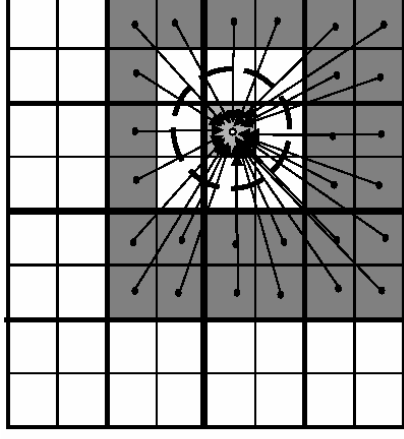


→ Level $l+1$ • · · → Level l

Downward pass



Level l



Level $l+1$

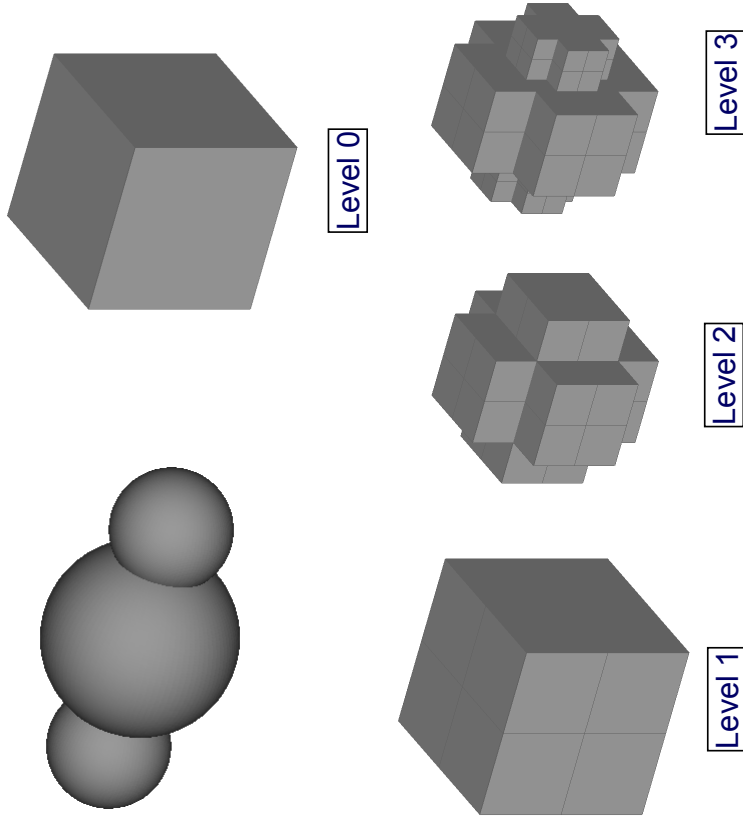
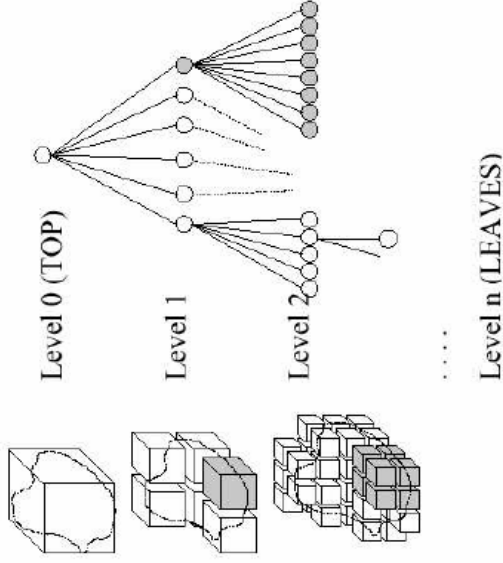
Multipole moments are accumulated from leaves to the root (**Upward pass**); and local moments are distributed from the root to the leaves (**Downward pass**). This is accomplished at a linear complexity.



Standard oct-tree

➤ **Algorithm**

➤ **One example**

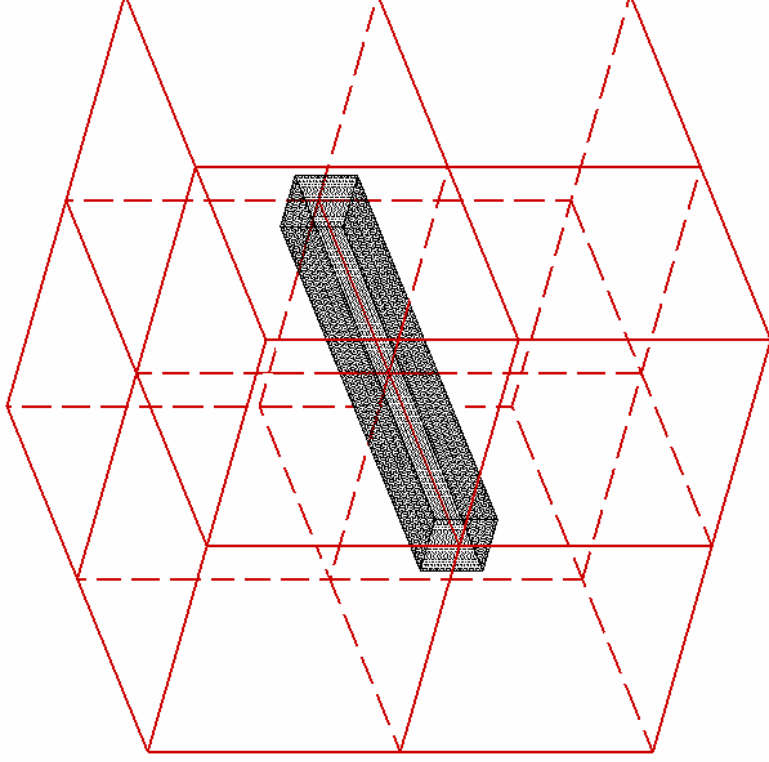




Standard oct-tree (2)

➤ Shortcoming

- Does not reflect the geometry of the computational domain!
- Resulted in a large number of M2L translations!

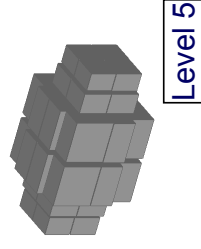
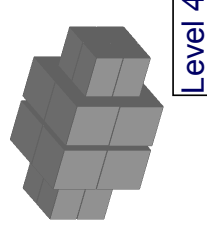
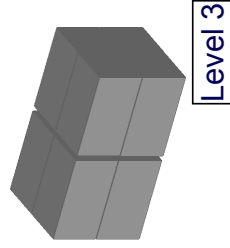
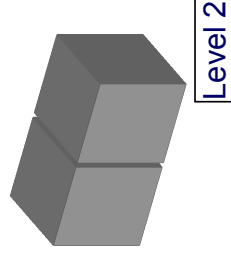
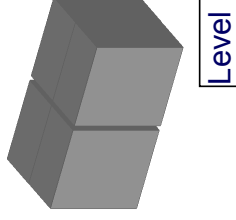
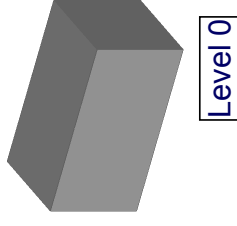
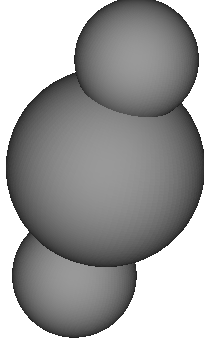




Binary tree

➤ **Differs from the standard tree:** ➤ **One example**

- Use rectangular boxes instead of cubes
- Subdivide a box in the longest direction
- Tighten the child boxes at each subdivision step



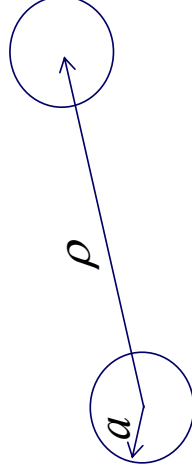
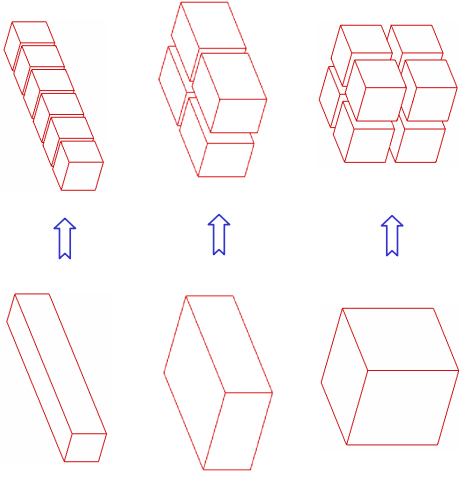


Adaptive tree

➤ **Differs from the standard tree:**

- Use rectangular boxes instead of cubes
- Subdivide a box according to the shape of the box
- Tighten the child boxes at each subdivision step
- Generalize the Downward Pass algorithm to allow M2L among the child boxes of a same parent box
- Determine the number of expansion terms in M2L by

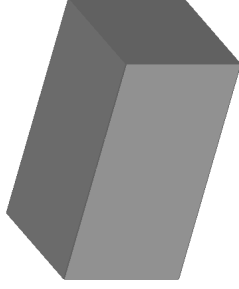
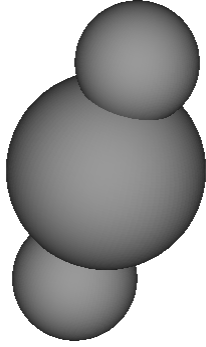
$$p = 0.117 p_{norm} / \log\left(\frac{a}{\rho - a}\right)$$



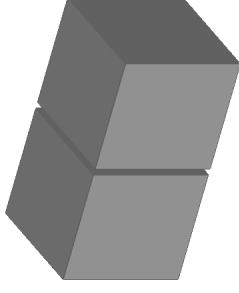


Adaptive tree (2)

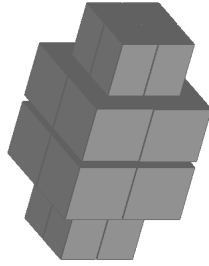
➤ **One example**



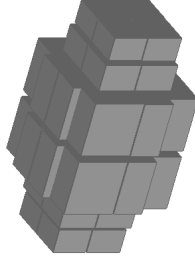
Level 0



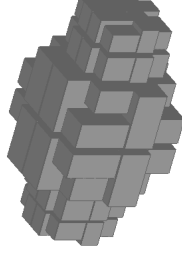
Level 1



Level 2



Level 3



Level 4



Test problems

Computer: **desktop computer with an Intel(R) Pentium(R) 4 CPU (1.99GHz)**

Analytical solution: $\phi = x^3 + y^3 + z^3 - 3yx^2 - 3xz^2 - 3zy^2$

Maximum nodes in a leaf: **60**

Number of expansion terms: $p = 10$

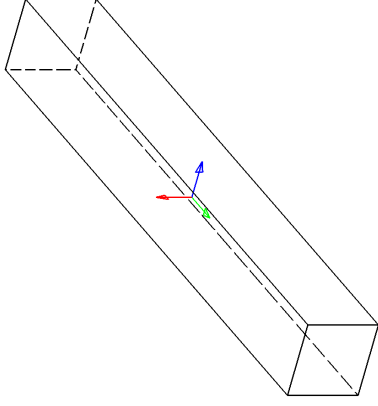
Iterative solver: **GMRES**

Convergence criterion: **relative error $< 10^{-5}$**

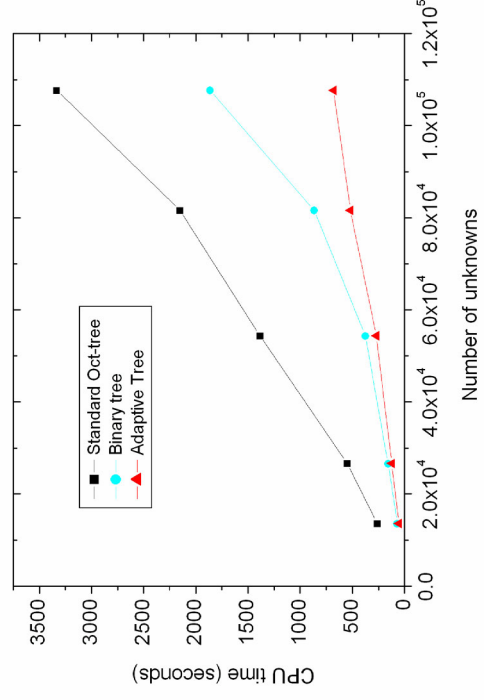
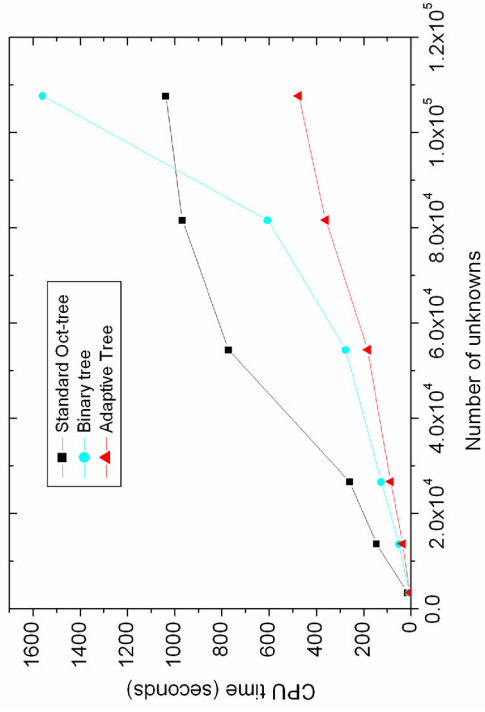
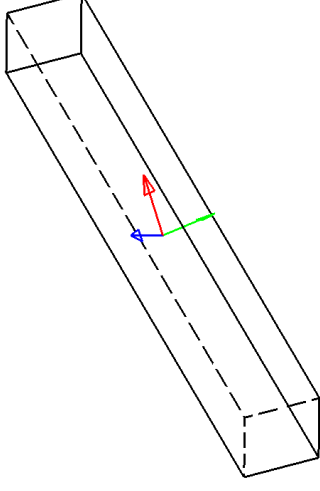


Test problems (2)

➤ A slender box

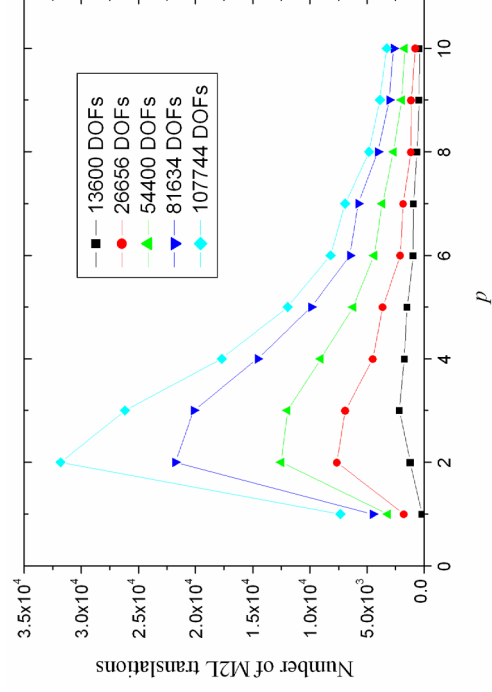
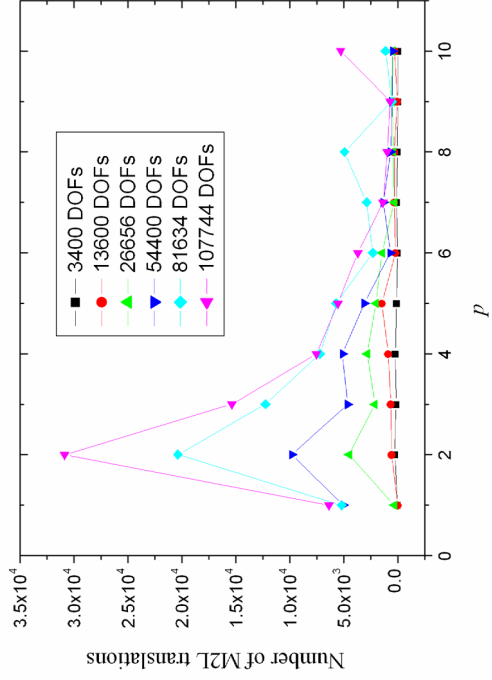
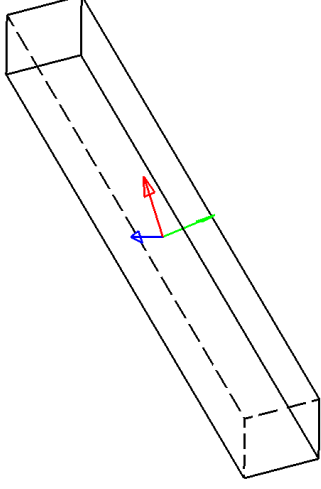
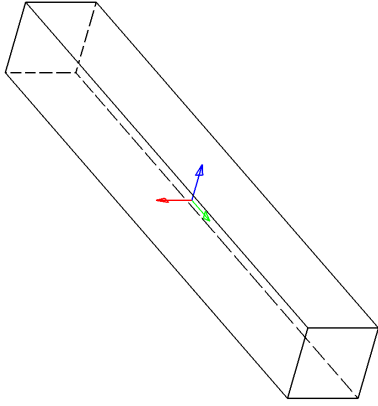


Dimensions: 20x20x160





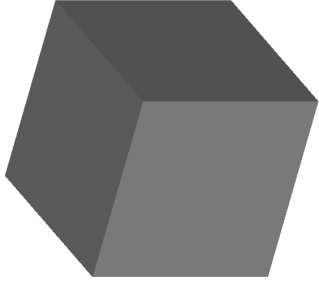
Test problems (3)



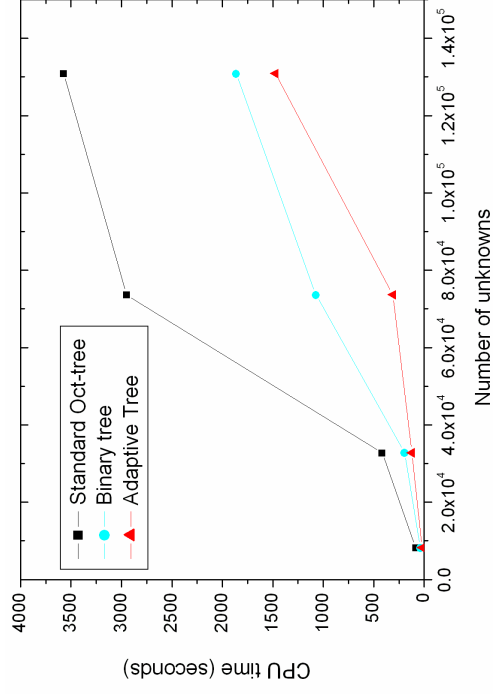
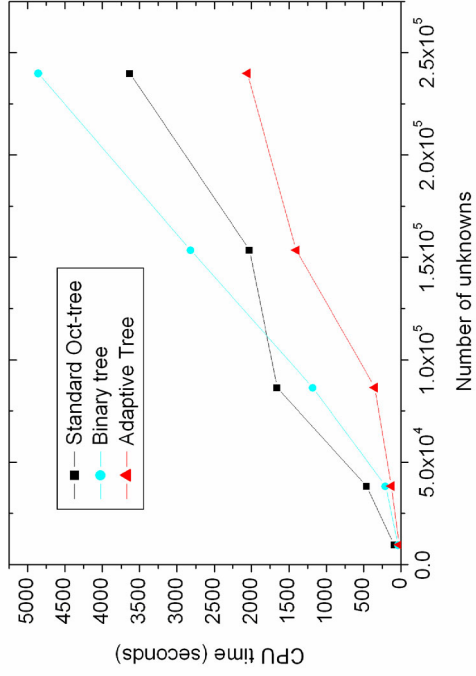
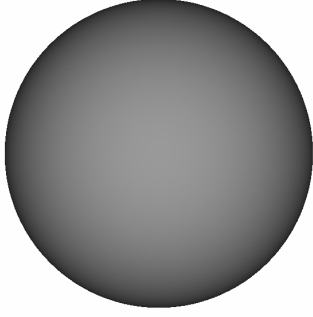


Test problems (4)

➤ **A cube**



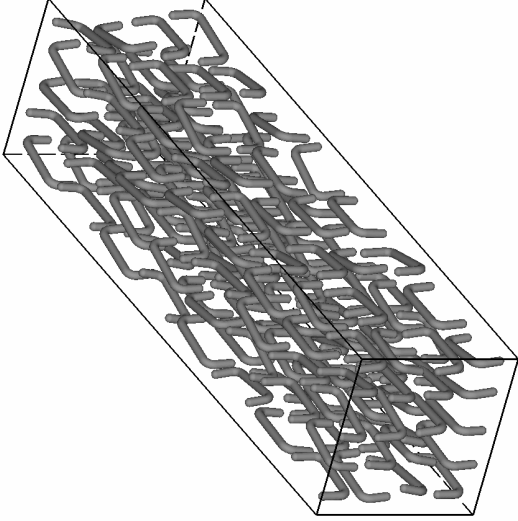
➤ **A sphere**



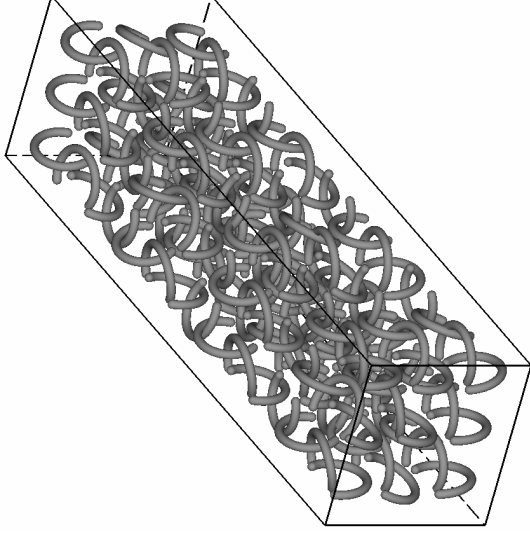


Test problems (5)

➤ CNT composite simulation



	κ	Nodes	Time (s)
Oct-tree	1.356	165153	58656
Adaptive tree	1.337	165153	9776



	κ	Nodes	Time (s)
Oct-tree	0.917	109314	32378
Adaptive tree	0.904	109314	5396



Conclusion remarks

- An adaptive tree data structure has been proposed. The new tree data structure is more flexible in matching the geometry (global and local) of the computational domain.
- Moreover, an adaptive value for the number of terms of the truncated series for M2L translations is used. This value is determined by the distance between the two interacting cells.
- Numerical examples show that the adaptive algorithm leads to trees with more compact cells and shallow depth, and runs significantly faster than the standard oct-tree.