A super cool scientific Title ...

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Abstract

Simulating large pairwise interactions is a very important issue for Scientific research. It plays an important role in Astrophysics to know the dynamics of galaxies, in plasma physics or in our case in biophysics. This kind of simulations is typically with a complexity of $\mathcal{O}(N^2)$ which scales badly with the size of the system.

Some other techniques, such as the PME (Particle Mesh Ewald) and the FMM (Fast Multipole Method) are able to obtain a complexity of respectively $\mathcal{O}(N\log(N))$ and $\mathcal{O}(N)$.

La simulation de larges systemes de particules en interaction est tres importante pour le calcul scientifique. Elles jouent un role important en Astrophysique pour connaître la dynamique des galaxies, en physique des plasmas ainsi que, dans notre cas, en Biophysique.

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Chapter 1

Presentation of the Lab

Max Planck Institute for Biophysical Chemistry

The Max-Planck Institute for Biophysical Chemistry is part of the Max-Planck Institues, which are an ensemble of research center throughout Germany (As the CNRS in France could be).

The Max-Planck Institute for Biophysical Chemistry, located in Göttingen, is one of the biggest research centers in the Biophysics field in Europe. Its research fields includes Molecular Dynamics Simulation, but also NMR or Biochemistry. The departments are : (from http://www.mpibpc.mpg.de/groups)

- Patrick Cramer Molecular Biology
- Gregor Eichele Genes and Behavior
- Dirk Görlich Cellular Logistics
- Christian Griesinger NMR-based Structural Biology
- Helmut Grubmüller Theoretical and Computational Biophysics
- Stefan W. Hell NanoBiophotonics
- Herbert Jäckle Molecular Developmental Biology
- Reinhard Jahn Neurobiology
- Reinhard Lührmann Cellular Biochemistry
- Marina V. Rodnina Physical Biochemistry
- Melina Schuh Meiosis
- Alec M. Wodtke Dynamics at Surfaces

Department of Theoretical and Computational Biophysics

This departments led by both Helmut grubmüller and Bert de Groot aims at understanding of the physics and function of proteins, protein complexes, and other biomolecular structures at the atomic level. For this purpose, complex computer simulations of the atomistic dynamics are carried out.

There is basically two main parts in the lab, one part is on really studying systems in order to know their dynamics (for example the case of the ribosome, which is studied in the lab), and another part which is more focusing on the methods, regardless of the biological system:

Chapter 2

Context of the Internship

The Context of this Internship is driven by the "SPPEXA (Software for exascale computing) / GromEx" project funded by the DFG (Deutsche Forschungsgeimeinschaft). The Idea of this project is to create a flexible and fast solver for computing forces and potentials, which is a preliminary for molecular simulations.

A poster¹ of the project can be found below :

 $^{^{1}} from\ http://www.mpibpc.mpg.de/grubmueller/sppexa$





German Priority Programme 1648 "Software for Exascale Computing"

SPPEXA - Findings & Goals SPPEXA - Chronology Massive parallelism (on- and cross-chip) 2006: discussion in the German Research Foundation (DFG) on the necessity of a requires fundamentally new concepts funding initiative for HPC software Not "racks without brains", but software is the key to this paradigm shift 2010: initiative out of German HPC commu-Fundamental research (→DFG), in contrast to nity, referring to increasing activities on HPC software elsewhere (USA: NSF, DOE; Japan; other (more application-oriented) initiatives (→ German Federal Ministry of E & R) China: G8) 2010: discussion with DFG's Executive Establish collaborative, interdisciplinary Committee, suggestion of a flexible, co-design of HPC applications and HPC strategically initiated SPP 2011: submission of the proposal, inter-national reviewing, and formal acceptance Focus on six research directions: Computational algorithms Application software 2012: Review of project sketches and full Programming System software proposals Data management and exploration Software tools SPPEXA – Implementation Two three-year funding phases Overall budget of 3,7 M € per year Funded via DFG's strategy fund Interdisciplinary consortia of 3-5 groups Consortia address at least two of SPPExa's six research directions SPPEXA - Current Status Two-stage application process with 68 sketches handed in, overall volume (1) sketches and (2) full proposals of 19 M € per year applied for Global strategic coordination, following 80 different universities, institutes and the established procedures of Collaborative companies represented by 240 national and 15 international PIs Research Centres (SFB) Close collaboration with respective inter- 24 sketches invited for full proposals national programmes intended ■ 13 full proposals accepted for funding Launch of programme and projects in January 2013 www.sppexa.de

Figure 2.1: Poster for the SPEXXA project

Currently the method used for computing electrostatic forces is called the PME (Particle Mesh Ewald). It works nicely but one of its problems is that the algorithms cannot br efficiently parallellised tasks as there is a lot of communication between the CPU and GPU cores. The idea would be to replace this method with a new method called the Fast-Multipole Method which is based on a tree Structure and may allow an greater parrallization of the system as wanted.

So the Idea of the Internship is first to know if the both method method have the same accuraracy and for which parameters.

Chapter 3

Methods for computing electrostatic forces

The electrostatic potential of a group of N of charge q_i at position r_i is defined as:

$$V_C = \frac{1}{4\pi\epsilon_0} \sum_{i=1}^{N} \sum_{j>i} \frac{q_i q_j}{r_{ij}}$$
 (3.1)

Computing this potential is very important in different areas of physics, such as plasma physics, or in our case molecular dynamics where the forces between charged particles need to be computed in an efficient and scalable way, the being computed using

$$\vec{F}_i = -\vec{\nabla}_i V(\vec{r}) \tag{3.2}$$

3.1 $\mathcal{O}(N^2)$ method

In this section, we will explain the most basic method to compute pairwise interactions and explain why the method leads to long computation times and sometimes .

3.1.1 Naive Method

The coulombian interaction between two charged particles can be written the following way:

$$\overrightarrow{F}_{A\to B} = \frac{q_A q_B \hat{r}_{AB}}{4\pi\epsilon_0 |R_{AB}|^2} \tag{3.3}$$

where q_A and q_B are respectively the charges of A and B, and R_{AB} is the distance between A and B.

In the thesis we will simplify the units of (3.3) for computational reasons by just writing:

$$\overrightarrow{F}_{A\to B} = \frac{q_A q_B \hat{r}_{AB}}{|R_{AB}|^2} \tag{3.4}$$

3.1. $\mathcal{O}(N^2)$ method

The Corresponding potential for a charged particle, with a charge q is :

$$V = \frac{1}{r} \tag{3.5}$$

The first, naive way to compute electrostatic forces is the following: in order to compute the force acting on one particle, it is needed to obtain the coulombic interaction for each pair of particles.

So if we consider a set of N charged particles, N-1 interactions are needed to compute the force acting on one specific particle. So in order to know the forces of the set of particles, $N \cdot (N-1)$ operations are needed, hence an algorithmic complexity of $\mathcal{O}(N^2)$.

This gives the following algorithm:

```
\begin{array}{c} \textbf{input} \ : \text{A set of } N \text{ charged Particles} \\ \textbf{output: A List of the forces for each particle} \\ For \ each \ particle \ i; \\ \textbf{for } i \leftarrow 1 \ \textbf{to} \ N-1 \ \textbf{do} \\ & | \ add \ interaction \ between \ particle \ i \ and \ particle \ j; \\ \textbf{for } j \leftarrow i+1 \ \textbf{to} \ N \ \textbf{do} \\ & | \ \text{force} \ [i] \leftarrow \text{force} \ [i] + \text{computeForce} (i,j) \ ; \\ \textbf{end} \\ \textbf{end} \end{array}
```

Algorithm 1: Naive method

The complexity of such a computation limits its use to rather small systems and is not really usable for bigger systems such as proteins or astrophysical systems.

3.1.2 Possible improvements

A possible improvement is to limit the interaction to a certain radius : if the distance between two particles if greater than R_0 , then the force is set to 0.

So we have the following system:

$$\overrightarrow{F}_{A \to B} = \begin{cases} \frac{q_A q_B \hat{r}_{AB}}{|R_{AB}|^2} & \text{if } R_{AB} < R_0\\ \overrightarrow{0} & \text{otherwise} \end{cases}$$
 (3.6)

This technique is for example used for Lennard-Jones potentials $(V_{LJ} = 4\epsilon[(\frac{\sigma}{r})^1 2 - (\frac{\sigma}{r})^6])$, where the Intensity of the force is quickly decreasing. It allows to limit the number of interactions to only the close neighbors.

However, one of the problems of this optimisation, especially for long-range interactions such as coulombic interactions is that using a cut-off can lead to artefacts: A particle feels the force, then crosses the cut-off radius. Suddenly, the particle doesn't feel any force anymore, thus the artifacts as it is showed figure 3.1.

3.2. Fourier Transform-Based methods

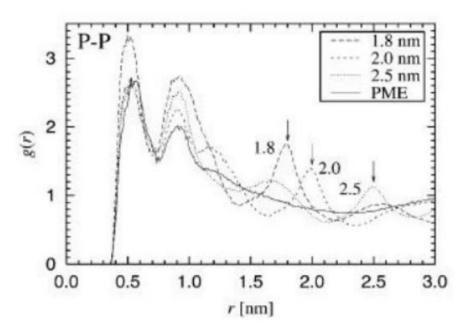


Figure 3.1: Radial distribution function (RDF) g(r) between the two central atoms in the head-group of a molecule: Cutoff distances are indicated by arrows. from [ADD REFERENCE]

As we can see in figure 3.1, the radial distribution of the distance between two atom shows a peak, corresponding to the cutoff of the system. This shows that by using a cut-off technique we might see some artifacts.

So the two reasons we don't use a $\mathcal{O}(N^2)$ method is first because of its algorithmic complexity, and using some optimization techniques can also lead to artifacts that are detrimental to the accuracy of the simulation.

3.2 Fourier Transform-Based methods

In this section, we will explain techniques using periodic boundary conditions and Fourier transformation in order to compute the potentials and the forces of the particles.

3.2.1 Ewald Summation

This subset of techniques comes from a theoretical physics technique called the Ewald summation.

Using periodic boundary conditions, the potential V of one particle of the system is:

$$V = \sum_{n_x, n_y, n_z} \sum_{i}^{N} \sum_{j}^{N} \frac{q_i q_j}{r_{ij}}$$
 (3.7)

3.2. Fourier Transform-Based methods

where n_x, n_y, n_z are the box index vector.

The equation (3.7) is conditionally convergent and slow to converge. One technique discovered by Ewald is to split the potential in two absolutely convergent terms and one constant term:

If the system is neutral, ie. $\sum_{i=1}^{N} q_i = 0$, The idea is to split the potential the following way:

$$V = \frac{1}{r} = \frac{f(x)}{r} + \frac{1 - f(x)}{r} \tag{3.8}$$

$$V = V_{\text{direct}} + V_{\text{reciprocal}} \tag{3.9}$$

We want to choose f, so that $\frac{f(x)}{r}$ is quickly decaying and $\frac{1-f(x)}{r}$ is as smooth as possible. In this decomposition $\frac{f(x)}{r}$ is as small as possible, even 0 above a certain cut-off. Then, $\frac{1-f(x)}{r}$ is smooth enough so we can compute its Fourier transform with just a few k vectors. This gives a quick computation of the reciprocal space.

A good choice is often $f(x)=\mathrm{erfc}(\alpha x)$, where α is called the splitting parameter, and $\mathrm{erfc}(x)=\frac{2}{\sqrt{\pi}}\int_x^{+\infty}e^{-t^2}\mathrm{d}t$

so we obtain the following potentials:

For the direct space:

$$V_{\text{direct}} = \frac{1}{2} \sum_{n_x, n_y, n_z} \sum_{i,j}^{N} q_i q_j \frac{\text{erfc}(\alpha r)}{r}$$
(3.10)

and for the reciprocal space :

$$V_{\text{reciprocal}} = \frac{1}{2\pi V} \sum_{i,j}^{N} q_i q_j \sum_{\substack{n_x^*, n_x^*, n_z^* \\ n_x^*, n_x^*}} \frac{\exp(-(\pi \overrightarrow{m}/\alpha)^2 + 2\pi i \overrightarrow{m} \cdot (r_i - r_j))}{m^2}$$
(3.11)

where n_x^*, n_y^*, n_z^* are the vectors in the reciprocal space.

Once the electrostatic potentials are obtained, it is possible to differentiate the potentials to obtain the force on each particle.

let p = (x, y, z) a coordinate of the system, then :

$$\overrightarrow{F}_{p}^{\text{direct}} = \overrightarrow{\nabla}_{p} V_{\text{direct}} \tag{3.12}$$

so we have:

$$\overrightarrow{F}_{p}^{\text{direct}} = q_{i} \sum_{i=1, i \neq j}^{N} \sum_{\mathbf{n}} q_{j} \frac{(r_{ij,n})_{p}}{r_{ij,n}^{3}} \left\{ \operatorname{erfc}(\alpha(r_{ij,n}) + \frac{2\alpha}{\sqrt{\pi}} r_{ij,n} \exp(-(\alpha r_{ij,n})^{2}) \right\}$$
(3.13)

3.2. Fourier Transform-Based methods

$$\overrightarrow{F}_{p}^{\text{reciprocal}} = \frac{2q_i}{L} \sum_{i=1, i \neq j}^{N} \sum_{n^* \neq 0} \frac{n_p^*}{n^{*2}} \exp\left(-\left(\frac{\pi m}{\alpha L}\right)^2\right) \sin\frac{2\pi}{L} m \cdot r_{ij}$$
(3.14)

Equations 3.13 and 3.14 can then be used to compute the force on each particle by adding the direct space and the reciprocal space contribution.

3.2.2 PME

One way to improve this method is to compute the Fourier sum using the FFT (Fast Fourier Transform). This allows the algorithm to get a complexity of $\mathcal{O}(N\log N)$. There is different methods that are based on the Ewald summation, namely the **P3M** (Particle-Particle Particle-Mesh Method) or the **FFP** (Fast Fourier Poisson method).

In our case, we will focus on the ${\bf PME}$ as it is the method currently used in GROMACS.

We remind that the potential for the reciprocal space $V_{\text{reciprocal}}$ is written in equation (3.11):

$$V_{\text{reciprocal}} = \frac{1}{2\pi V} \sum_{i,j}^{N} q_i q_j \sum_{\mathbf{n}^*} \frac{\exp{-(\pi \overrightarrow{m}/\alpha)^2} + 2\pi i \overrightarrow{m} \cdot (r_i - r_j)}{m^2}$$

This equation can be rewritten as:

$$V_{\text{reciprocal}} = \frac{1}{2\pi V} \sum_{\mathbf{n}^* \neq 0}^{N} \frac{\exp{-(\pi \overrightarrow{m}/\alpha)^2}}{m^2} S(-\mathbf{n}^*) S(\mathbf{n}^*)$$

where $S(\mathbf{n}^*)$ is defined as the Structure factor:

$$S(\mathbf{n}^*) = \sum_{k=1}^{N} q_k \exp(2\pi i \mathbf{n}^* \cdot \mathbf{r})$$
(3.15)

The idea of the method is the following: In order to apply a FFT to the system, it is needed to "map" the positions of the charges on a grid of size $p \times p$. We will define the size of the Grid as the Fourier Spacing. Let also define $\mathcal{F}(Q)$ the 3D FFT of Q. The charges q_i are mapped to the grid using interpolation. Originally, Lagrange interpolation was used, but now a b-spline interpolation is used. The order of interpolation is called the PME Order

then the Structure factor can be approximated as its FFT:

$$S(\mathbf{m}) \approx \widetilde{S}(\mathbf{m}) = \mathcal{F}(Q)(\mathbf{m})$$
 (3.16)

so the reciprocal energy can also be approximated by:

$$V_{\text{reciprocal}} \approx \widetilde{V}_{\text{reciprocal}} = \frac{1}{2\pi V} \sum_{\mathbf{n}^* \neq 0}^{N} \frac{\exp(-(\pi \overrightarrow{m}/\alpha)^2)}{m^2} \mathcal{F}(Q)(\mathbf{m}) \mathcal{F}(Q)(-\mathbf{m}) \quad (3.17)$$

It can be shown that the complexity of such a system is $\mathcal{O}(N \log(N))$ which is a much bigger improvement compared to the $\mathcal{O}(N^2)$ complexity of the direct algorithm. However, one of its disadvantages is that it relies on a periodic sum, which is theoretically infinite. Furthermore, the algorithm doesn't scale well for large scale parallelism[REFERENCE?], hence the need for a more scalable algorithm for the computations of electrostatic forces.

3.3 Fast Summation methods

In the previous section we showed a method which allows to improve the speed of the computation. In this section, another algorithm will be explained, which has a complexity of $\mathcal{O}(N)$.

3.3.1 Mathematical preliminaries

Let's move back to the potential created by a charged particle as in (3.5),

$$V = \frac{1}{d}$$

Let two particles $A(a, \alpha, \beta)$ and $B(r, \theta, \phi)$, separated by a certain distance d: The distance between the particles is d = |a - r|. We would like to achieve is to "factorize the addition", having a product of one function depending only on a and one depending only on r.

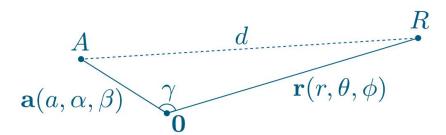


Figure 3.2: Figure representing two particles $A(a, \alpha, \beta)$ and $B(r, \theta, \phi)$, separated by a certain distance d

The inverse distance can be written as:

$$\frac{1}{d} = \frac{1}{|a-r|} = \frac{1}{\sqrt{a^2 + r^2}}$$

This distance can then written as a series:

$$\frac{1}{d} = \frac{1}{\sqrt{a^2 + r^2}} = \sum_{l=0}^{+\infty} P_l(u)\mu^l$$
 (3.18)

where $\mu = \frac{a}{r}$ and $u = \cos \gamma$. The $P_l(u)$ are the Legendre polynomials of degree l.

with some manipulation as explained [ADD REFERENCE], we obtain :

$$\frac{1}{|r-a|} = \sum_{l=0}^{+\infty} \sum_{m=-l}^{+l} \frac{(l-m)!}{(l+m)!} \frac{a^l}{r^{l+1}} P_{lm} \cos(\alpha) P_{lm} \cos(\theta) e^{-im(\beta-\alpha)}$$
(3.19)

in order to approximate the scheme, we can truncate the series to a certain order p, we call this order the **multipole order**

$$\frac{1}{|r-a|} \simeq \sum_{l=0}^{\mathbf{p}} \sum_{m=-l}^{+l} \frac{(l-m)!}{(l+m)!} \frac{a^l}{r^{l+1}} P_{lm} \cos(\alpha) P_{lm} \cos(\theta) e^{-im(\beta-\alpha)}$$
(3.20)

We can now rewrite the summation the following way:

$$\frac{1}{|r-a|} \simeq \sum_{l=0}^{p} \sum_{m=-l}^{+l} \underbrace{\frac{a^{l}}{(l+m)!} P_{lm}(\cos(\alpha)) e^{-im\beta}}_{O_{lm}(\mathbf{a})} \underbrace{\frac{(l-m)!}{r^{l+1}} P_{lm}(\cos(\theta)) e^{+im\phi}}_{M_{lm}(\mathbf{r})}$$

$$(3.21)$$

hence,

$$\frac{1}{|r-a|} \simeq \sum_{l=0}^{p} \sum_{m=-l}^{+l} O_{lm}(\mathbf{a}) M_{lm}(\mathbf{r})$$
(3.22)

We have shown that it is possible to "factorize" the inverse of the the distance $\frac{1}{d}$ so we can obtain two independent terms for two independent particles $O_{lm}(\mathbf{a})$ and $M_{lm}(\mathbf{r})$

We can then define the multipole moment $\omega_{lm}(q, \mathbf{a}) = qO_{lm}(\mathbf{a})$ and the Taylor-like moment $\mu_{lm}(q, \mathbf{r}) = qM_{lm}(\mathbf{r})$

Then, it is possible to show that it is possible to obtain the following "bipolar expansion" for two particles associated with two different origins:

$$\frac{1}{|\mathbf{a}_1 - \mathbf{a}_2 + \mathbf{R}|} = \sum_{l=0}^{+\infty} \sum_{j=0}^{+\infty} \sum_{m=-l}^{+l} \sum_{k=-j}^{+j} (-1)^j \cdot O_{lm}(\mathbf{a}_1) \cdot M_{l+j,m+k}(\mathbf{R}) \cdot O_{jk}(\mathbf{a}_2)$$
(3.23)

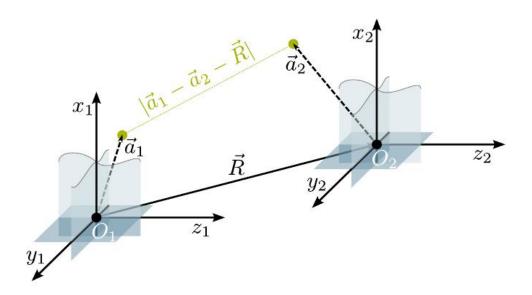


Figure 3.3: Scheme of two particles expanded according to two different origins [from ADD REFERENCE]

3.3.2 Workflow of the algorithm

Once some Mathematical preliminaries are set, it is possible to explain the workflow of the Algorithm.

3.3.2.1 Spliting the Space

The first step of the scheme is to split the space in order to generate different groups. The idea is to recursively split the space in eight octants, created the so-called structure of an *octree*. So the number of boxes is 8^{D-1} , where D is the depth of the oct-tree.

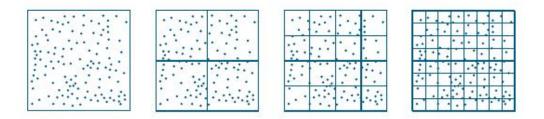


Figure 3.4: Example of a division of the space in boxes. The following example is in 2D so the structure of the space is a *quadtree*.

The number of subdivisions is called the $depth\ D$. For example, in figure 3.5, we can observe from left to right the depths 0,1,2 and 3.

It is then possible to assign each particle of the simulation to the boxes of lowest depth.

3.3.2.2 Defining a Separation Criterion

For the further Computations it is needed to define a separation criterion ws: It is the ws^{th} next neighbors of a given box. Two boxes A and B are called *next neighbors* is they are at the same level and box B is enclosed by a box of size $(2ws+1)^3$ around the center of A

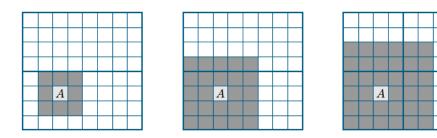


Figure 3.5: Example of the influence of the separation criterion ws for a given box A, from left to right, ws = 1, 2, 3.

This separation criterion divides the space into two parts. If two particles are close so they are in the same grey space as defined figure (3.8), then, they will interact via a direct $\mathcal{O}(N^2)$ coulombic interaction; else they will interact via using the FMM Method.

It is most of the time not possible to set the separation criterion ws to 0, as the multipole expansion might overlap. Using a bigger ws means that the order of the multipole expansions need to be less important for the same precision, however this will increase the number of direct $\mathcal{O}(N^2)$ operation, which are more computationally expensive.

3.3.2.3 PASS 1: Computing the Multipole moments

Once every particle is assigned to its box, we will compute the multipole moment ω_{lm}^j of each box.

$$\omega_{lm}^{j}(q, \mathbf{a}) = q^{j} a_{j}^{l} \widetilde{P}_{lm}(\cos(\alpha_{j})) e^{-im\beta_{j}}$$
(3.24)

Then each multipole order can be moved to the Center of the box using the so-called M2M (Multipole to Multipole) operator:

$$\omega(\mathbf{a} + \mathbf{b}) = \sum_{j=0}^{k} \sum_{k=-j}^{j} \omega_{jk}(\mathbf{a}) O_{l-j,m-k}(\mathbf{b})$$
(3.25)

 $O_{l-j,m-k}(\mathbf{b})$ is called the M2M operator

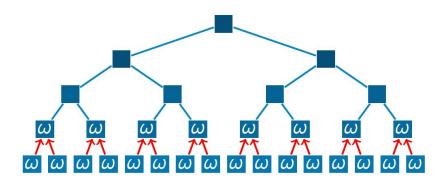


Figure 3.6: Example in 1-D showing the shifting of the multipole moments using the M2M operator.

3.3.2.4 PASS 2: Transforming distant expansions into multipoles

Now it is possible to transform distant expansions (above the separation criterion) ω_{lm} into a Taylor-like moment μ_{lm} . This transformation is done over at most 189 boxes. This transformation is applied for each depth of the tree.

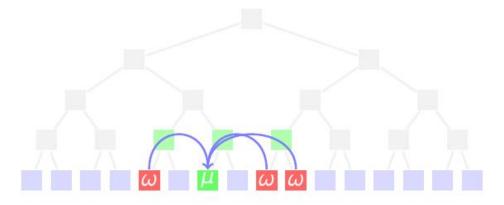
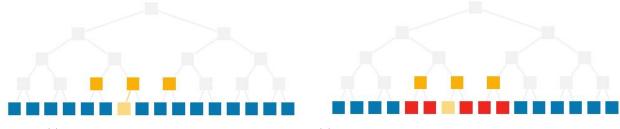


Figure 3.7: Example in 1-D showing the transformation of distant expansions into multipoles $\frac{1}{2}$

In order to compute this transformation, two different rules have to be applied:

- The first rule is to take the non-separated parent boxes, as showed subfigure (a) below.
- Then the children of the parent box are selected, taking into account the separation criterion for the child boxes.



(a) First, the parents of the box are chosen

(b) Then, then children of the parents are chosen. Then the separation criterion can apply

Figure 3.8: Interation Rules for the transformation of the multipole expansions

Mathematically speaking, the transformation from multipole expansions to multipole moments is given using the M2L (Multipole to Local) Operator

$$\mu_{lm}(\mathbf{b} - \mathbf{a}) = \sum_{j=0}^{+\infty} \sum_{k=-j}^{j} M_{j+l,k+m}(\mathbf{b}) \cdot \omega_{jk}((\mathbf{a}))$$
 (3.26)

where $M_{j+l,k+m}(\mathbf{b})$ is the M2L operator: It allows to exchange, as figure (??) shows, information between boxes of the same level.

3.3.2.5 PASS 3: Shift Taylor-Like expansions down the tree

Once the local μ_{lm} expansions are computed, the expansion have to shifted to the lowest level of the tree using the L2L (Local to Local) operator.

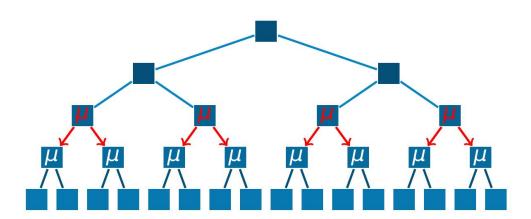


Figure 3.9: Example in 1-D showing the shifting of the local moments using the L2L operator.

This shift operation is done by the following formula:

$$\mu_{lm}(\mathbf{r} - \mathbf{b}) = \sum_{j=l}^{p} \sum_{k=-j}^{j} O_{j-l,k-m}(\mathbf{b}) \cdot \mu_{jk}((\mathbf{a}))$$
(3.27)

where $O_{j-l,k-m}(\mathbf{b})$ is the L2L operator : It allows to exchange, as figure (??) shows, information between levels downwards.

3.3.2.6 PASS 4: Computing forces and Energies

After PASS 3, we have at the lowest level both the multipole expansion as well as the local expansion.

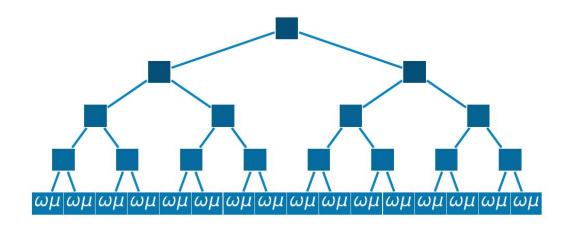


Figure 3.10: Situation after PASS 3

Then we have everything to compute the far-field contribution for the forces or the potentials of the system.

So we have for the Electric Field:

$$E_{FF} = \sum_{l=0}^{p} \sum_{m=-l}^{+l} \mu_{lm}(\mathbf{r})\omega_{lm}(\mathbf{a})$$
(3.28)

For the Potential:

$$\Phi_{FF} = -\sum_{l=0}^{p} \sum_{m=-l}^{+l} \mu_{lm}(\mathbf{r}) \nabla_{\mathbf{a}_i} [a_i^l \widetilde{P}_{lm}(\cos \alpha_i) e^{-im\beta_i}]$$
 (3.29)

and for the Force :

$$\mathbf{F}_{FF} = -\sum_{l=0}^{p} \sum_{m=-l}^{+l} \mu_{lm}(\mathbf{r}) \nabla_{\mathbf{a}_i} \left[a_i^l \widetilde{P}_{lm}(\cos \alpha_i) e^{-im\beta_i} \right]$$
 (3.30)

Then the near-field contribution can be easily computed:

Chapter 4

Comparing FMM and PME accuracy

In this chapter I will more precfisely explain my work done in the Lab, which consisted in comparing the PME methods, which is for example used by gromacs, and the FMM mnethod, used by a solver made between the Julich Forschungszentrum and the Max Planck Institute for Biophysics in Göttingen.

In will first explain for Gromacs work and then

4.1 Presentation of GROMACS

GROMACS (GROningen MAchine for Chemical Simulations) is a simulation software used for Molecular dynamics. It was originally developed in the Biophysical Chemistry department of University of Groningen but it is now developed arround the world. GROMACS is made to be as fast as possible using all possible techniques to improve its performance (MPI, GPU computing)

4.1.1 Structure of a File

The program is used by modifying text files that giving information on the structure of the system that has to be simulated and the parameters of the simulation. The files are the following

- The *.pdb file are the *Protein DataBank* file: In this file is denoted the position and the type of the particles of the system. It also gives the size of the simulation box
- The *.top files (namelt topoloy files) are the file where the properites of the atoms are defined. These properties are for example the charge, the Van-der-Waals parameters or the binding forces of the system. This configuration is often done with force field files such as AMBER or CHARMM
- The *.mdp files are defining the physical and computational parameters of the simulation. It is defining the methods used for the simulation. In our case we are mostly interested in the electrostatics part of the computation

4.2. Presentation of fmsolvr

; it is then possible to select the method (PME or Cut-Off methods) and the parameters for the ${\rm PME}$

4.1.2 PME parameters

In this paragraph I will explain the parameters it is possible to play with:

CutOff The first parameter is the CutOff: It allows to set the Cut-Off radius for the Cut-Off method, but it also sets the difference between the direct part and the reciprocal part in the PME method

Fourier Spacing The other important parameter is the fourier spacing. I remind that in the PME method a 3D FFT is done in order to compute the energies and the forces of the system. The FFT is so computed on a grid: The dimension of the grid is given by the Fourier Spacing parameter.

PME Order The PME order gives the order of the interpolation. For instance, PMEorder= 4 corresponds to a cubic interpolation.

4.1.3 Command to launch a GROMACS simulation

The workflow to laumch a GROMACS Simulation is the following:

First generate a .tpr binary file conaining all the information about the simulation. The file is processed using the *grompp* (Gromacs Preprocessor). The command is :

```
grompp -f mdpfile
  -p topFile.top
  -c pdbFile.pdb
  -o tprFile.tpr
```

where the input files are the *.mdp , *.top and *.pdb (Respectively properties of the Simulation, of the atom, and the opsition of the At and contains arroundoms). The *.tpr file ris the output file.

It is then possible to run the simulation using the program mdrun, which as its name stands, runs the md simulation. A possible example is:

```
mdrun -s tprfile.tpr
-ntomp = 4
-ntmpi = 4
-nsteps= 0
```

Where the .tpr File is now the input file, containing everything about the simulation, -nsteps is the number of time steps needed (In the case of this Intership, no time intergration is needed as just the forces are required). It is also possible to choose the number of CPU cores needed for the simulation (Using MPI and OpenMP).

4.2 Presentation of fmsolvr

The software used to compute the electrostatic forces with the FMM is not gromacs (at least not for the moment). It is developed by both the Max-

4.3. Making GROMACS and fmsolvr comparable

Planck Institut for Biophysical Chemistry and the Jülich Forschungszentrum . In the following chapters, we will call this code fmsolvr.

The codebase is mostly written in C++. There exists several git branches for the program, allowing different versions of the system. The First version is a sequential version, which is the "basic" version of the FMM code . There also exists a version which allows periodic boundary conditions : We will use this version a lot as we need to campare it to the PME, which, by construction, uses periodic boundary conditions.

The input file, is a *.hpp file containing 4 arrays of size N, where N is the number of atoms in the simulation : one charge array q, and three array x,y,z for the positions of the atoms.

4.2.1 FMM Parameters

The simulation can be launched using the following command:

DEPTH=\$DEPTH MULTIPOLEORDER=\$MULTIPOLEORDER WS=\$WS

OPENBOUNDARY=O UNITBOX=\$UNITBOX CENTER=c

./fmmtest \$inputFile.qxyz \$outputFile.dat

The environment variables are the following:

DEPTH The maximal subdivision of the space : if DEPTH=n the space will be divided in 8^n boxes.

 $\mathbf{MULTIPOLEORDER}$ The Order of truncation in a the series, for instance :

$$\frac{1}{|r-a|} \simeq \sum_{l=0}^{p} \sum_{m=-l}^{+l} \frac{(l-m)!}{(l+m)!} \frac{a^{l}}{r^{l+1}} P_{lm} \cos{(\alpha)} P_{lm} \cos{(\theta)} e^{-im(\beta-\alpha)}$$

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where the multipole order is the red p in the first summation.

WS This gives the separation criterion $ws \geq 1$, which separates the far-field computation from the near space.

OPENBOUDARY if OPENBOUNDARY = 0, periodic boundary conditions will be used, if OPENBOUNDARY=1, then openbondaries will be used.

UNITBOX Gives the size of the simulation box (in nm.

4.3 Making GROMACS and fmsolvr comparable

The first part, if we want to make the PME and the FMM comparable, is to use the same simulation for both programs, hence requiring coding some tools transforming a GROMACS-compatible file to fmsolvr-compatible one. Then, we saw that the results we obtained weren't good enough as one system was using some dipole correction, so we needed to implements the dipole correction to the FMM system.

4.3.1 File manipulation

In this section will be explained the file modifications needed in order to have the simulations comparable on both systems.

4.3.1.1 Making GROMACS only compute electrostatics forces

In this Internship we just want to compute the electrostatic forces on our system. However, what we obtain in the GROMACS output file is the sum of all forces on each atom. These forces may include forces such as for instance Lennard-Jones potentials. $(V_{LJ} = 4\epsilon[(\frac{\sigma}{r})^12 - (\frac{\sigma}{r})^6])$.

So the first thing to do is to modify the topology file (*.top), so there is no more Lennard Jones potential at all; this snippet is added to the *.top file:

```
[ atomtypes ]
```

```
type atnum mass charge ptype sigma epsilon CLA 17 35.450000 0.000 A 0.000000000000 0.00000 SDD 11 22.989770 0.000 A 0.000000000000 0.00000
```

In the expression of the potential, we set $\epsilon = 0$ and $\sigma = 0$, so it gives $V_{LJ} = 0$. So we have modified the files such that there is only the electrostatic interaction at play in the simulation.

4.3.1.2 Assure compatibility between *.qxyz and *.pdb files

The next thing to do is to have the same simulation system for both methods: So I needed to make a few scripts to transform a *.pdb file in a *.qxyz or a *.hpp file.

```
The scripts are written in python, the workflow is the following: *.hpp \stackrel{\text{qxyz2hpp.py}}{\longleftarrow} *\text{qxyz} \stackrel{\text{qxyz2gromacs.py}}{\longrightarrow} *.\text{pdb}
```

So it is now possible to move the positions of the atoms one software to another software. (PME to FMM or FMM to PME)

4.3.1.3 Output positions and forces to a Text file for GROMACS

The last preliminary thing to do is to make the GROMACS Simulation print the positions and the forces in an easy and detailled way. We decided to modify the Gromacs code to add after the so-called do_force(...) function, which computes the forces for the system, a routine that prints the force as well as the positions in order to be easily used afterwards.

4.4 First Comparisions and Dipole Correction

4.4.1 Silicamelt Simulation

The first thing we did afterwards it to take a good precision for both methods and compare the forces.

The model we are using is called the "Silicamelt" system which is a system composed of SiO_2 atoms: It is a system compased of about 100000 atoms, the silicon atoms (Si) have a +2.4e charge and the Oxygen has a charge of -1.2e. The size of the simulation box is 124.120 nm.

The parameters for the PME simulation are cutoff=1.2nm, PME order=12 and fourier spacing=0.015 nm.

The parameters for the FMM are : Depth=4, Multipole Order=40, and WS(Separation Criterion) =4. The simulations is done with periodic boundaries

The relative error is computed the following way, assuming the PME is more accurate than the FMM:

$$error_{\text{relative}} = \frac{F_{FMM} * F_{PME}}{F_{PME}}$$
 (4.1)

Then the following histogram can be plotted:

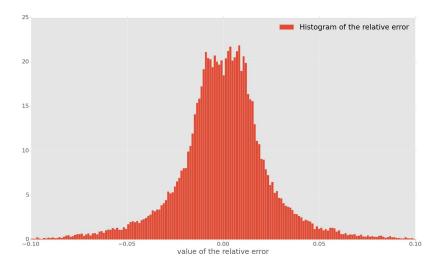


Figure 4.1: Histogram representing the relative error for a silicamelt simulation : The histogram is computed before any diple correction.

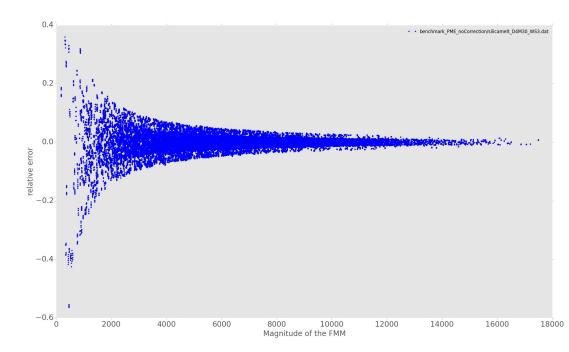


Figure 4.2: Figure representing the Magnitude of the force in the x-axis and the error in relative error in the y-axis. It is possible to see that the biggest errors are when the magnitude of the force is smaller.

Some statistical information about the distribution gives us a maximum error of about 56.35%, this maximum error is much bigger than expected and is therefore not acceptable.

If we study the distribution of errors according to the magnitude of system we can see that the biggest errors are some smaller magnitudes.

4.4.2 Dipole correction

4.4.2.1 Study of a two-particle system

In order to understand these big errors, we first decided to study how 2 particles behave. Then if the errors are small enough, it is possible to move to bigger systems.

We also observed that if we set 4 particles in a square in a quadrupole, so there is no dipole here, the value of the FMM system and the PME system are matching

The hypothesis is that the dipole moment may be corrected in one system and not in another system. So the idea is to take two particles, one positively charged and the other negatively charged; the two particles will form a dipole, namely:

$$\vec{p} = q\vec{d} \tag{4.2}$$

where \vec{d} is the displacement vector pointing from the negative charge to the positive charge.

Then we can vary the distance between the two particles, compute the force using both the FMM method and the PME method and compare the force.

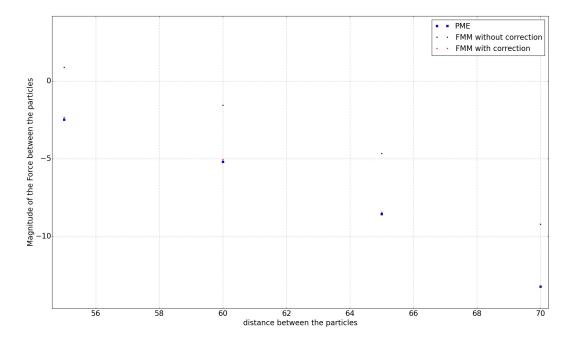


Figure 4.3: Figure representing the distance between two particles the x-axis (here 55,60,65,70 nm) and the Magnitude of the force for PME (Blue Square), FMM without correction (Black point) and with correction(red star). It is possible to see that the dipole correction makes the FMM magnitude much closer to the PME.

The dipole correction is done the following way :

It is possible to modify the potential in the FMM code using the following equation:

$$\nabla_i E_{\text{after correction}} = \nabla_i E_{textbeforecorrection} - \frac{4\pi}{3 \cdot V} q_i \mathbf{d}$$
 (4.3)

where V is the volume of the simulation box, q_i the charge of the particle and **d** is the dipole moment.

The dipole moment can be computed from the multipole moments, which contains the dipole moment : The multipole moments are contained in a triangluar array (because of symmetry and memory space reasons) the following way :

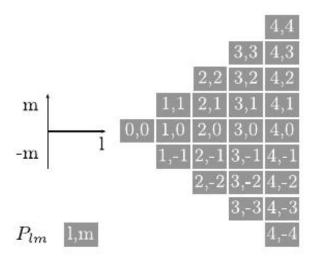


Figure 4.4: Structure of the array containing the multipoles $\omega_{lm}\mathbf{a}$): the 'm' values are on the y-axis and the 'l'-values are on the x-axis. The dipole moment is contained in the boxes (1,1), (1,0) and (1,-1)

As showed in figure (4.4), it is possible to obtain the dipole moment from the multipole ones, using the following formula:

$$\begin{cases}
\vec{d}_x = -2 * \Re(\omega_{1,1}) \\
\vec{d}_y = 2 * \Im(\omega_{1,1}) \\
\vec{d}_z = -\Re(\omega_{1,0})
\end{cases}$$
(4.4)

Then the correction can be applied using equation (4.3)

The code used for this correction, in C++11, is the following:

```
for (size_t i = 0; i < n; ++i) {
    auto cell_volume = (abc.a).x * (abc.b).y * (abc.c).z; // Only works
        with square Cell Unit
    std::cout << "Periodic Vector = " << (abc.a).x << std::endl;
    std::cout << "Volume " << cell_volume << std::endl;

std::cout << Real3(ordered_particles[i]) << "\t" <<reference_center
        << std::endl;

auto dr = Real3(ordered_particles[i]) - reference_center;
    std::cout << "dr = " << dr << std::endl;

const auto& omega = tree_omega[0][0];
    dipole_moment.x = -omega.get(1,1).real()*2;
    dipole_moment.y = omega.get(1,1).imag()*2;</pre>
```

```
dipole_moment.z = -omega.get(1,0).real();
// Potential Correction
auto correction = (4.0*PI/(3.0 * cell_volume )) * (dr.x *
    dipole_moment.x) + (dr.y * dipole_moment.y) + (dr.z *
    dipole_moment.z);
std::cout << "Potential Before = " << potential[i] << std::endl;</pre>
std::cout << "Correction = " << correction << std::endl;</pre>
potential[i] -= correction;
std::cout << "Potential After = " << potential[i] << std::endl;</pre>
efield[i].x -= 4*PI/3 * dipole_moment.x;
efield[i].y -= 4*PI/3 * dipole_moment.y;
efield[i].z -= 4*PI/3 * dipole_moment.z;
        Real q = -(-ordered_particles[i].s); // negative gradient
        force[i] = efield[i] * q - dipole_moment *
            (4.*PI/(3.0*cell_volume)) * q;
   Ec2 += q * potential[i];
    }
    Ec += Ec2 * 0.5;
```

4.4.2.2 Back to the silicamelt simulation with dipole correction

Now we can relaunch the simulation with the same parameters as before, including the dipole correction. We obtain now a maximum error of 3.52%, which is a much bigger improvement compared to the 56.35% of the former simulation.

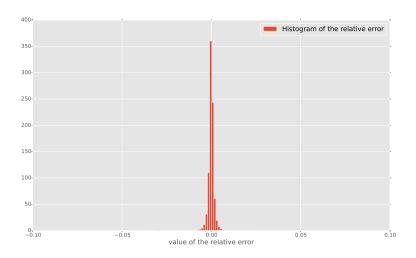


Figure 4.5: Histogram representing the relative error for a silicamelt simulation: The histogram is computed with the dipole correction. The scale in x-direction is the same as in figure (??)

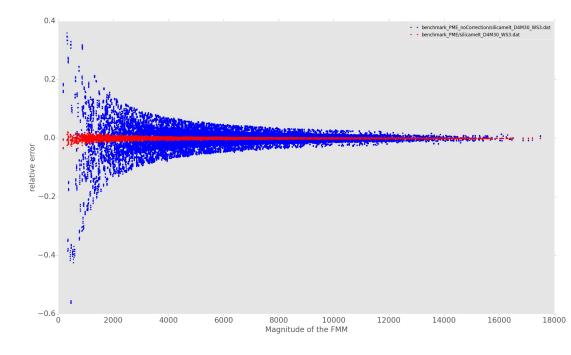


Figure 4.6: Figure representing the Magnitude of the force in the x-axis and the error in relative error in the y-axis. The blue points represent the force before the dipole correction, then red points after the dipole correction

4.5. Comparing with an analytical solution: the NaCl system

4.5 Comparing with an analytical solution : the NaCl system

One other thing we want to try is to see the convergence of the scheme. The idea is to know which set of parameters for the FMM has the same accuracy than a given set of parameters for the PME method.

In order to study that,

4.5.1 Generating the system

4.5.2 Some explainaitions on the Owl system

4.5.3 Error plots

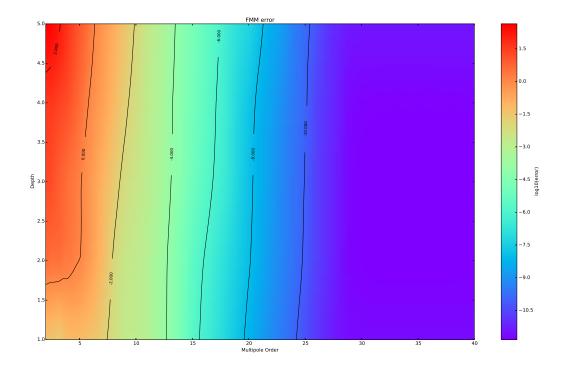


Figure 4.7: Figure representing the Magnitude of the force in the x-axis and the error in relative error in the y-axis. The blue points represent the force before the dipole correction, then red points after the dipole correction

Appendices

Appendix A

Glossary and Abbreviations

 ${\bf GROMACS}\,$ GROningen MAchine for Chemical Simulations

 ${f PME}$ Particle Mesh Ewald

 \mathbf{FMM} Fast Multipole Method

Owl MPI-BPC Cluster Used to make the Simulations

 \mathbf{MPI} Message Passing Interface

 \mathbf{Omp} OpenMP

Appendix B

Structure of Gromacs files

The contents...