

## LA-UR-17-25908

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Title: FleCSPH notes

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Intended for: Report

Issued: 2017-07-17

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# FleCSPH notes

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July 5, 2017

## 1 FleCSI

FleCSI is a compile-time configurable framework designed to support multi-physics application development. As such, FleCSI provides a very general set of infrastructure design patterns that can be specialized and extended to suit the needs of a broad variety of solver and data requirements. FleCSI currently supports multi-dimensional mesh topology, geometry, and adjacency information, as well as n-dimensional hashed-tree data structures, graph partitioning interfaces, and dependency closures.

FleCSI introduces a functional programming model with control, execution, and data abstractions that are consistent both with MPI and with state-of-the-art, task-based runtimes such as Legion and Charm++. The abstraction layer insulates developers from the underlying runtime, while allowing support for multiple runtime systems including conventional models like asynchronous MPI.

The intent is to provide developers with a concrete set of user-friendly programming tools that can be used now, while allowing flexibility in choosing runtime implementations and optimization that can be applied to future architectures and runtimes.

FleCSI's control and execution models provide formal nomenclature for describing poorly understood concepts such as kernels and tasks. FleCSI's data model provides a low-buy-in approach that makes it an attractive option for many application projects, as developers are not locked into particular layouts or data structure representations.

FleCSI currently provides a parallel but not distributed implementation of Binary, Quad and Oct-tree topology. This implementation is based on space filling curves domain decomposition, the Morton order.

The current FleCSI version requires the implementation of a *driver* and a *specialization\_driver*. The role of the *specialization\_driver* is to provide the data distribution. This feature is not complete in FleCSI code and we provide it. The next step will be to incorporate it directly from FleCSPH to FleCSI as we reach a good level of performance. Then the *driver* represent the general execution of the resolution without worrying of the data locality and communications. As FleCSI is an On-Development code the structure may change in the future and we keep track of these changes in FleCSPH.

## 2 Domain and tree construction

As explain in the previous section we described the FleCSI framework. In this part we will give more details on the domain decomposition and the tree construction and search.

### 2.1 Domain decomposition

In the current version of FleCSI the domain decomposition is done using the Morton Ordering. It first allows us to describe, sort and distribute the particles based on a unique value.

Several kind of space filling curves can be use:

- Morton order: easy to compute, interlace the bits of X, Y and Z positions to create the key. The problem of this space filling curve are the discontinuities.
- Hilbert-Peano: More complex to compute, interlace the bits like the Morton order but add rotations base on the gray-code. but solve the problem of discontinuities.
- Other space-filling curves: Hexagonal space filling curves, ...?

This first implementation is based on the Morton ordering which is used during several steps:

- The distribution part, to be able to split the particles between the processes providing a good locality in the data.
- The tree construction and search. The keys are used

### 3 Binary, Quad and Oct - trees

Describe the tree and the splitting version of it

Explain the ghosts, exclusive and shared particles.

### 4 General algorithm

The main distributed algorithm is presented in algorithm 1

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#### Algorithm 1 Main algorithm

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```

1: procedure SPECIALIZATION_DRIVER(input data file  $f$ )
2:   read data file in a distributed way
3:   Set physics constant from data file
4:   while iterations do
5:     Distribute the particles using distributed quick sort           ▷ Using Morton keys
6:     Compute total range
7:     Generate the local tree
8:     Share branches
9:     Compute the ghosts particles
10:    Update ghosts data
11:    Do physics
12:    Update ghosts data
13:    Do physics
14:    Distributed output to file
15:  end while
16: end procedure

```

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In the current version the 5 is base on a distributed quick sort algorithm. Each process send to the master (or submaster for larger cases) a sample of its keys. We determined this size to be 256 Ko of key data per process but can be refine for larger simulations. Then the master determines the general ordering for all the processes and shares the pivots. Then each process locally sort its local keys and, in a global communication step, the particles are distributed to the process on which they belong. This distribution provide a quick distribution but can lead to bad load balancing.

- The ordering can be not perfect in term of number of particles per processes. But by changing the number of data exchanged to the master can lead to better affectation.
- The load balance also depend on the number of neighbors of each particles. If a particle get affected a poor area with large space between the particles this can lead to bad load balancing too.

After the sorting step the local tree can be created on each process. To be able to look for the ghosts and shared particles we need to share some information with the neighbors on the tree. In the 8 line we will search for these interesting branches. We compute the global bounding box of each processes and based on this information each process can then compute the interesting branches to share from its local tree. These new information are then added in the local tree by considering *NON\_LOCAL* particles. This data structure does just contain the position and mass of the distant particle.

The branch sharing allow to compute the ghosts for this step. Each process performs a local search in the tree and computes the required ghosts particles (the *NON\_LOCAL* bodies). Those data for shared and ghosts are stored

and are use to share the complete particle information when 10 is invoke. As the ghosts data remain the same within an iteration, the 10 can be use several to update local information on remote particles.

## 5 I/O

In large simulation an efficient, parallel and distributed I/O implementation is require. We base this first implementation on HDF5 file structure with H5Part and H5Hut. The I/O code was developed internally in the LANL and provide a simple way to write and read the data in H5Part format. The first requirement is to allow the user to work directly with the Paraview visualization tool and splash<sup>1</sup>.

## 6 SPH Formulation in FleCSPH

Smoothed particle hydrodynamics (SPH) is explicit numerical mesh-free method that solves hydrodynamical partial differential equations (PDE) by discretizing in set of fluid elements called particle. The fundamental SPH formulation is

$$\langle A \rangle(\vec{r}) \simeq \sum_b \frac{m_b}{\rho_b} A(\vec{r}_b) W(|\vec{r} - \vec{r}_b|, h) \quad (1)$$

where  $W$  is the smoothing kernel,  $h$  is the smoothing length (hydro interaction range) that evolved for each particle.

SPH has several advantages. SPH can handle deformations, low densities, and vacuum very well. Also, it conserves mass, linear and angular momentums, and energy by its construction that implies independent of the numerical resolution. Another strong benefit of using SPH is its exact advection of fluid properties. Furthermore, the particle structure of SPH easily combines with tree method for solving Newtonian gravity through N-body simulations. However, there are several cons of using SPH. It is restricted to low-order convergence. Also, SPH requires careful setup of initial distribution of particles. Further, it can be struggle to resolve turbulence dominated flows and special care must be taken when handling high gradients such as shocks and surface structure of neutron stars

Here, we want to solve the Lagrangian conservation equations for mass, energy and momentum of an ideal fluid such that

$$\frac{d\rho}{dt} = -\rho \nabla \cdot \vec{v} \quad (2)$$

$$\frac{du}{dt} = \left( \frac{P}{\rho^2} \right) \frac{d\rho}{dt} \quad (3)$$

$$\frac{d\vec{v}}{dt} = -\frac{\nabla P}{\rho} \quad (4)$$

where  $d/dt = \partial_t + \vec{v} \cdot \nabla$

By using the volume element  $V_b = m_b/\rho_b$ , we can formulate the Newtonian SPH scheme such that

$$\rho_a = \sum_b m_b W_{ab}(h_a) \quad (5)$$

$$\frac{du_a}{dt} = \frac{P_a}{\rho_a^2} \sum_b m_b \vec{v}_{ab} \cdot \nabla_a W_{ab} \quad (6)$$

$$\frac{d\vec{v}_a}{dt} = -\sum_b m_b \left( \frac{P_a}{\rho_a^2} + \frac{P_b}{\rho_b^2} \right) \nabla_a W_{ab} \quad (7)$$

where  $W_{ab} = W(|\vec{r}_a - \vec{r}_b|, h)$ . Further, we add artificial viscosity (or artificial dissipation) terms in SPH formulation

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<sup>1</sup>Describe it! url too

such that

$$\left(\frac{du_a}{dt}\right)_{art} = \frac{1}{2} \sum_b m_b \Pi_{ab} \vec{v}_{ab} \cdot \nabla_a W_{ab} \quad (8)$$

$$\left(\frac{d\vec{v}_a}{dt}\right)_{art} = - \sum_b m_b \Pi_{ab} \nabla_a W_{ab} \quad (9)$$

$$(10)$$

In general, we can express the equations for internal energy and acceleration with artificial viscosity

$$\frac{du_a}{dt} = \sum_b m_b \left( \frac{P_a}{\rho_a^2} + \frac{\Pi_{ab}}{2} \right) \vec{v}_{ab} \cdot \nabla_a W_{ab} \quad (11)$$

$$\frac{d\vec{v}_a}{dt} = - \sum_b m_b \left( \frac{P_a}{\rho_a^2} + \frac{P_b}{\rho_b^2} + \Pi_{ab} \right) \nabla_a W_{ab} \quad (12)$$

$\Pi_{ab}$  may define different way but here we use

$$\Pi_{ab} = \begin{cases} \frac{-\alpha \bar{c}_{ab} \mu_{ab} + \beta \mu_{ab}^2}{\bar{\rho}_{ab}} & \text{for } \vec{r}_{ab} \cdot \vec{v}_{ab} < 0 \\ 0 & \text{otherwise} \end{cases} \quad \text{where} \quad \mu_{ab} = \frac{\bar{h}_{ab} \vec{r}_{ab} \cdot \vec{v}_{ab}}{r_{ab}^2 + \epsilon \bar{h}_{ab}^2} \quad (13)$$

For speed of sound,  $c_s$  the usual form is

$$c_s = \sqrt{\frac{\partial p}{\partial \rho}} \quad (14)$$

For example, from the Newton-Laplace equation,  $c = \sqrt{\frac{K_s}{\rho}}$  where  $K_s$  is a coefficient of stiffness, the isentropic bulk modulus. The values of  $\epsilon$ ,  $\alpha$ , and  $\beta$  can be chosen differently. Here, we use  $\epsilon = 0.01h^2$ ,  $\alpha = 1.0$ , and  $\beta = 2.0$ . Now, we need to test some cases for performance of code

## 6.1 Kernel

There are many kernels for SPH. We use simple cubic spline kernel for our case. Other higher order kernels will be added soon. The Monaghan's cubic spline kernel is

$$W(\vec{r}, h) = \frac{\sigma}{h^D} \begin{cases} 1 - \frac{3}{2}q^2 + \frac{3}{4} & \text{if } 0 \leq q \leq 1 \\ \frac{1}{4}(1-q)^3 & \text{if } 1 \leq q \leq 2 \\ 0 & \text{otherwise} \end{cases} \quad (15)$$

where  $q = r/h$ ,  $D$  is the number of dimensions and  $\sigma$  is a normalization constant with the values

$$\sigma = \begin{cases} \frac{2}{3} & \text{for 1D} \\ \frac{10}{7\pi} & \text{for 2D} \\ \frac{1}{\pi} & \text{for 3D} \end{cases} \quad (16)$$

Using this, we can calculate kernel gradient. Below procedure shows the steps of kernel gradient in 3D

$$W = \frac{1}{\pi h^3} \times \begin{cases} 1 - \frac{3}{2}(\frac{r}{h})^2 + \frac{3}{4}(\frac{r}{h})^3, & \text{si } 0 \leq \frac{r}{h} < 1 \\ \frac{1}{4}[2 - \frac{r}{h}]^3, & \text{si } 1 \leq \frac{r}{h} < 2 \\ 0, & \text{si } \frac{r}{h} \geq 2 \end{cases}$$

$$\text{And } r = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2 + (z_i - z_j)^2} \text{ with } r = \sqrt{u^2 + v^2 + w^2} \text{ and } \vec{r}_{ij} = \begin{cases} u = x_i - x_j \\ v = y_i - y_j \\ w = z_i - z_j \end{cases}$$

$$\begin{aligned}\vec{\nabla} \cdot W &= \frac{\partial W}{\partial u} \vec{x} + \frac{\partial W}{\partial v} \vec{y} + \frac{\partial W}{\partial w} \vec{z} = \frac{\partial W}{\partial r} \frac{\partial r}{\partial u} \vec{x} + \frac{\partial W}{\partial r} \frac{\partial r}{\partial v} \vec{y} + \frac{\partial W}{\partial r} \frac{\partial r}{\partial w} \vec{z} \\ &= \frac{\partial W}{\partial r} \left[ \frac{\partial r}{\partial u} \vec{x} + \frac{\partial r}{\partial v} \vec{y} + \frac{\partial r}{\partial w} \vec{z} \right] = \frac{\partial W}{\partial r} \frac{r_{ij}}{r}\end{aligned}$$

For  $0 \leq \frac{r}{h} < 1$  :

$$\frac{\partial W}{\partial r} = -\frac{3}{h^2}r + \frac{9}{4h^3}r^2$$

$$\vec{\nabla}_i W(r_{ij}, h) = \left( -\frac{3}{h^2}r + \frac{9}{4h^3}r^2 \right) \frac{r_{ij}}{r} = \left( -\frac{3}{h^2} + \frac{9}{4h^3}r \right) r_{ij}$$

For  $1 \leq \frac{r}{h} < 2$  :

$$\frac{\partial W}{\partial r} = \frac{-3}{4h} \left( 2 - \frac{r}{h} \right)^2 = \frac{-3}{4h} \left( 4 - \frac{4r}{h} + \frac{r^2}{h^2} \right) = \frac{-3}{h} + \frac{3r}{h^2} + \frac{-3r^2}{4h^3}$$

$$\vec{\nabla}_i W(r_{ij}, h) = \left( \frac{-3}{h} + \frac{3r}{h^2} + \frac{-3r^2}{4h^3} \right) \frac{r_{ij}}{r} = \left( \frac{-3}{hr} + \frac{3}{h^2} + \frac{-3r}{4h^3} \right) r_{ij}$$

So:

$$\vec{\nabla}_i W(r_{ij}, h) = \frac{1}{\pi h^4} \times \begin{cases} \left( -\frac{3}{h} + \frac{9}{4h^3}r \right) r_{ij}, & \text{si } 0 \leq \frac{r}{h} < 1 \\ \left( \frac{-3}{r} + \frac{3}{h} + \frac{-3r}{4h^2} \right) r_{ij}, & \text{si } 1 \leq \frac{r}{h} < 2 \\ 0, & \text{si } \frac{r}{h} \geq 2 \end{cases}$$

## 7 Applications

### 7.1 Sod Shock Tube

The Sod shock tube is the test consists of a one-dimensional Riemann problem with the following initial parameters

$$(\rho, v, p)_{t=0} = \begin{cases} (1.0, 0.0, 1.0) & \text{if } 0 < x \leq 0.5 \\ (0.125, 0.0, 0.1) & \text{if } 0.5 < x < 1.0 \end{cases} \quad (17)$$

This link shows some references and values for Sod shock tube problem that also includes boundary and jump conditions (<http://www.phys.lsu.edu/~tohline/PHYS7412/sod.html>).

Also, we would like to re-generate the shock test result from Rosswog's paper. In that paper, he shows the result of a 2D relativistic shock tube test where the left state is given by  $[P, v_x, v_y, N]_L = [40/3, 0, 0, 10]$  and the right state by  $[P, v_x, v_y, N]_R = [10^{-6}, 0, 0, 1]$  with  $\Gamma = 5/3$

In our code, we use below parameters to get results

### 7.2 Sedov Blast Wave

A blast wave is the pressure and flow resulting from the deposition of a large amount of energy in a small very localized volume. This is another great test problem for computational fluid dynamics field.

There are different version of blast wave test but we consider the analytic solution for a point explosion is given by Sedov, making the assumption that the atmospheric pressure relative to the pressure insider the explosion negligible. The position of the shock as a function of time  $t$ , relative to the initiation of the explosion, is given by

$$R(t) = \left( \frac{et^2}{\rho_0} \right)^{\frac{1}{\delta+2}} \quad (18)$$

with  $\delta = 2$  and  $\delta = 3$  for cylindrical and spherical geometry respectively. The initial density  $\rho_0$  whereas  $e$  is a dimensionless energy. Right behind the shock we have the following properties

$$\rho_2 = \frac{\Gamma+1}{\Gamma-1} \rho_0 P_2 = \frac{2}{\Gamma+1} \rho_0 w^2 v_2 = \frac{2}{\Gamma+1} w \quad (19)$$

where the shock velocity is

$$w(t) = \frac{dR}{dt} = \frac{2}{\delta + 2} \frac{R(t)}{t} \quad (20)$$

In numerical simulations, energy deposition in a single point is difficult to achieve. A solution to the problem is to make use of the bursting balloon analogue. Rather than depositing the total energy in a single point, the energy is released into a balloon of finite volume  $V$

$$e = \frac{(P - P_0)V}{\Gamma - 1} \quad (21)$$

The energy release in a balloon of radius  $r_0$  raises the pressure to the value

$$P = \frac{3(\Gamma - 1)e}{(\delta + 1)\pi r_0^\delta} \quad (22)$$

Here, we test 2D blast wave test. In this simulation, we use ideal gas EOS with  $\Gamma = 5/3$  and we are assuming that the undistributed area is at rest with a pressure  $P_0 = 1.0^{-5}$ . The density is constant  $\rho_0$ , also in the pressurized region.

### 7.3 Equations of State

To understand the inner property of stars, one needs to find the equation which describes the relation between the pressure of matter and its density, temperature and other compositions such that

$$P = P(\rho, T, Y_e, \dots) \quad (23)$$

First, we consider analytic equations of state that are relevant for binary neutron stars

#### 7.3.1 Ideal Gas

Ideal gas equation of state is

$$P(\rho, u) = (\Gamma - 1)\rho u \quad (24)$$

where  $\Gamma$  is the adiabatic index of the gas. For a monatomic gas, we set  $\Gamma = 5/3$ . For another test such as sod tube, people use  $\Gamma = 1.4$

#### 7.3.2 Piecewise Polytrope

For more relevant simulation, we choose piecewise polytropic EOS. (Ideal gas EOS is still good for many simple test cases like Sod tube) In our case, we assume constant entropy so that many thermodynamic situations can be approximated as polytropes or piecewise functions made up of polytropes.

For neutron star case, we assume degenerated Fermi gas of neutrons then polytropic constant for a non-relativistic degenerated neutron gas is

$$K_0 = \frac{(3\pi^2)^{2/3} \hbar^2}{5m_n^{8/3}} \quad (25)$$

where  $m_n$  is the mass of a proton and  $\hbar$  is a Planck constant. For polytropic index, we set

$$\gamma_0 = \frac{5}{3} \quad (26)$$

In the relativistic case,

$$\gamma_1 = \frac{5}{2} \quad (27)$$

Then, piecewise polytrope EOS is

$$P(\rho) = \begin{cases} K_0 \rho^{\gamma_0} & \text{if } \rho \leq \rho_0 \\ \frac{K_0 \rho_0^{\gamma_0}}{\rho_0^{\gamma_1}} \rho^{\gamma_1} & \text{if } \rho > \rho_0 \end{cases} \quad (28)$$

where  $\rho_0 = 5 \times 10^{14} \text{ g/cm}^3$ . We can combine the piecewise polytropic EOS with ideal gas to attain an EOS valid at both low and high densities. For more realistic studies, we need to consider different types of analytic EOSs such as Maxwell-Boltzmann and Helmholtz EOSs. Also, we will put the functionality that can control tabulated EOS.



### 7.3.3 Zero Temperature Equations of State

Another interesting problem using SPH is the double white dwarf (DWD) simulations for studying possible progenitors to type Ia supernovae. Here, we use zero temperature equations of state (ZTWD) as a variation of the self consistent field technique. In ZTWD, the electron degeneracy pressure  $P$  varies with the mass density  $\rho$  according to the relation

$$P = A \left[ x(2x^2 - 3)(x^2 + 1)^{1/2} + 3 \sinh^{-1} x \right] \quad (29)$$

where the dimensionless parameter

$$x \equiv \left( \frac{\rho}{B} \right)^{1/3} \quad (30)$$

and the constant A and B are

$$A \equiv \frac{\pi m_e^4 c^5}{3h^3} = 6.00288 \times 10^{22} \text{ dynes cm}^{-2} \quad (31)$$

$$\frac{B}{\mu_e} \equiv \frac{8\pi m_p}{3} \left( \frac{m_e c}{h} \right)^3 = 9.81011 \times 10^5 \text{ g cm}^{-3} \quad (32)$$

## 8 Time Integration Scheme

### 8.1 Leap-Frog Method

Leap-frog time integration is frequently used in particle simulation. The name comes that the velocities are updated on half steps and the positions on integer steps so the two leap over each other. After computing accelerations, one step takes the form

$$v^{i+1/2} = v^{i-1/2} + a^i \Delta t \quad (33)$$

$$r^{i+1} = r^i + v^{i+1/2} \Delta t \quad (34)$$

For  $v^{i+1}$ , a common approximation in SPH is to assume that the velocity at the current time plays a minor role in the computation of the acceleration (i.e. velocity changes are small with each time step) and then the following approximation can be made

$$v^{i+1} = \frac{1}{2}(v^{i-1/2} + v^{i+1/2}) \quad (35)$$

At the first step, we only have initial velocity so we need to follow below routing

$$v^{1/2} = v^0 + a^0 \Delta t / 2 \quad (36)$$

$$r^1 = r^0 + v^{1/2} \Delta t \quad (37)$$

The time step is adaptive and determined with  $\Delta t = \text{Min}(\Delta t_1, \Delta t_2)$ :

$$\Delta t_1 = k \text{Min}_i \left( \frac{h_i}{c_i + 1.2\alpha c_i + 1.2\beta \text{Max}_j \mu_{ij}} \right) \text{ with } k \approx 0.1$$

$$\Delta t_2 = \text{Min}_i \sqrt{\left( \frac{h_i}{|\vec{v}_i|} \right)}$$

## 9 Solving Lane-Emden Equation

We need to determine the density function based on the radius.

As we consider the star as a polytropic fluid, we use the equation of Lane-Emden which is a form of the Poisson equation:

$$\frac{d^2 \theta}{d\xi^2} + \frac{2}{\xi} \frac{d\theta}{d\xi} + \theta^n = 0 \quad (38)$$

With  $\xi$  and  $\theta$  two dimensionless variables. There is only exact solutions for a polytropic index  $n = 0.5, 1$  and  $2$ . In our work we use a polytropic index of  $1$  which can correspond to a NS simulation.

For  $n = 1$  the solution of equation 38 is:

$$\theta(\xi) = \frac{\sin(\xi)}{\xi} \quad (39)$$

We note  $\xi_1 = \pi$ , the first value of  $\xi$  as  $\theta(\xi) = 0$ .  $\theta(\xi)$  is also defined as:

$$\theta(\xi) = \left( \frac{\rho(\xi)}{\rho_c} \right)^{\frac{1}{n}} = \frac{\rho(\xi)}{\rho_c} \quad (40)$$

With  $\rho_c$  the internal density of the star and  $\rho$  the density at a determined radius.  $\xi$  is defined as:

$$\xi = Ar = \sqrt{\frac{4\pi G}{K(n+1)} \rho_c^{(n-1)/n}} \times r = \sqrt{\frac{2\pi G}{K}} \times r \text{ (for } n = 1\text{)}$$

With  $K$  a proportionality constant.

From the previous equations we can write the stellar radius  $R$  as:

$$R = \sqrt{\frac{K(n+1)}{4\pi G}} \rho_c^{(1-n)/2} \xi_1 = \sqrt{\frac{K}{2\pi G}} \times \xi_1 \quad (41)$$

(We note that for  $n = 1$  the radius does not depend of the central density.)

If, for example, we use dimensionless units as  $G = R = M = 1$  (for the other results we use CGS with  $G = 6.674 \times 10^{-8} \text{ cm}^3 \text{ g}^{-1} \text{ s}^{-2}$ ) We can compute  $K$  as:

$$K = \frac{R^2 2\pi G}{\xi_1^2} \quad (42)$$

	$NS_1$	$NS_2$	$NS_3$	$NS_4$
Radius (cm)	$R = G = M = 1$	1500000	1400000	960000
K	0.636619	95598.00	83576.48	39156.94

Then we deduce the density function of  $r$  as :

$$\rho(\xi) = \frac{\sin(A \times r)}{A \times r} \times \rho_c \text{ with } A = \sqrt{\frac{2\pi G}{K}}$$

As we know the total Mass  $M$ , the radius  $R$  and the gravitational constant  $G$  we can compute the central density as:

$$\rho_c = \frac{MA^3}{4\pi(\sin(AR) - AR\cos(AR))}$$

Then we normalize the results to fit  $R = M = G = 1$ :  $K' = K/(R^2 G)$ ,  $m'_i = m_i/M$ ,  $h'_i = h_i/R$ ,  $\vec{x}'_i = \vec{x}_i/R$

## 9.1 Gravitational force

The self-gravity  $\vec{F}_i^{Grav}$  for each particles:

$$\vec{F}_i^{Grav} = \sum_j G \frac{m_i m_j}{(|\vec{r}_j - \vec{r}_i|)^3} \vec{r}_{ij} \quad (43)$$

In this part we will need Fast Multiple Method (FMM) for the computation to avoid  $O(N^2)$  complexity.

## 10 Resolution order

The resolution is done in this order:

```

Load data from file
while TotalTime not reached do
  Move particles
  Apply rotation with defined angular velocity
  Compute density  $\rho$ 
  Compute pressure  $P$  and sound speed  $c$ 
  Compute  $\vec{F}_{hydro}$ 
  Compute  $\vec{F}_{grav}$ 
  if relaxation then
    Compute  $\vec{F}_{roche}$  or  $\vec{F}_{rot}$ 
  end if
  Compute acceleration  $\vec{a}$  with the equation of motion
  Compute  $\Delta t$ 
  Compute new velocity from new acceleration
  if output step then
    output data to file
  end if
  TotalTime  $\leftarrow$  TotalTime +  $\Delta t$ 
end while

```

## 11 Hydrostatic Equilibrium

The initial data are based on a cubic lattice within a sphere of radius  $R$ . The density function, based on radius,  $\rho(\vec{r})$  is known using the result of the Lane-Emden equation ( we use polytropic index  $n = 1$  here). The mass associate to each particle  $i$  of the total  $N$  particles:

$$m_i = \frac{\rho(\vec{r}_i)}{n_r} \text{ with } n_r = \frac{3N}{4\pi R^3}$$

The smoothing length is define constant and the same for all particles for all the simulation:

$$h = \frac{1}{2} \sqrt{\frac{3N_N}{4\pi n}}$$

Here we choose  $N_N$ , the average number of neighbors, to be 100.

### 11.1 Roche lobe problem

To create Hydrostatic Equilibrium Models we use a different equation of motion. This version use Roche Lobe:

$$\frac{d\vec{v}_i}{dt} = \frac{\vec{F}_i^{Grav}}{m_i} + \frac{\vec{F}_i^{Hydro}}{m_i} + \vec{F}_i^{Roche} - \frac{\vec{v}_i}{t_{relax}} \quad (44)$$

With  $t_{relax} \leq t_{osc} \sim (G\rho)^{-1/2}$  and where  $\vec{F}_i^{Roche}$  is:

$$\vec{F}_i^{Roche} = \mu(3+q)x_i\hat{x} + \mu q y_i\hat{y} - \mu z_i\hat{z}$$

With  $\mu$  to be determined (for us  $\mu = 0.069$ ) and  $q = \frac{M'}{M} = 1$  as the two polytropes have the same total mass. This is apply to each star to get the equilibrium and the simulate the tidal effect.

## 11.2 Darwin problem

This is the way we use to generate the final simulation. The equation of motion for the relaxation is now:

$$\frac{d\vec{v}_i}{dt} = \frac{\vec{F}_i^{Grav}}{m_i} + \frac{\vec{F}_i^{Hydro}}{m_i} + \vec{F}_i^{Rot} - \frac{\vec{v}_i}{t_{relax}} \quad (45)$$

With  $t_{relax}$  same as before and  $\vec{F}_{Rot}$  defined by:

$$\vec{F}_{Rot} = \Omega^2(x_i\vec{\hat{x}} + y_i\vec{\hat{y}}) \quad (46)$$

With  $\Omega = \sqrt{\frac{G(M+M')}{a^3}}$ .

Or  $L_z = Q_{zz}\Omega$  and  $Q_{zz} = \sum_i(x_i^2 + y_i^2)$ . At  $t = 0$  we compute the total angular momentum  $L_z$  which stay constant. Using it during the relaxation we can compute  $\Omega$  as:  $\Omega = \frac{L_z}{Q_{zz}}$  just by recomputing  $Q_{zz}$ .

Here the scheme is in  $N^2$  but just for the relaxation step.

For this relaxation we use two stars generated as before, applying equation of motion 45. Using  $a$  as the distance between the two polytropes (Here  $a = 2.9$  for  $R = 1$ ) and  $\vec{\hat{x}}$  going for the center of the first to the second star, and  $\vec{\hat{z}}$  is like the rotation vector.

## 12 Gravitation computation

The algorithm is presented in algorithm 2. In this algorithm, the *macangle* is the angle of the Multipole Acceptance Criterion. The acceleration is, for a center of mass  $c$ , the sum of contributions from the local particles and the distant cells with:

$$\vec{f}_c(\vec{r}_c) = - \sum_p \frac{m_p(\vec{r}_c - \vec{r}_p)}{|\vec{r}_c - \vec{r}_p|^3} \quad (47)$$

With  $p$  the particle inside this cell and *cell* the cells that are accepted with the MAC. Here we directly consider the gravitational acceleration, we don't take in account the mass of the center of mass  $c$ . And  $G = 1$  in our context.

The acceleration at a point from this center of mass is based on taylor series:

$$\vec{f}(\vec{r}) = \vec{f}_c(\vec{r}_c) + \left\| \frac{\partial \vec{f}_c}{\partial \vec{r}} \right\| \cdot (\vec{r} - \vec{r}_c) + \frac{1}{2}(\vec{r} - \vec{r}_c)^\top \cdot \left\| \frac{\partial \vec{f}_c}{\partial \vec{r}_c \partial \vec{r}_c} \right\| \cdot (\vec{r} - \vec{r}_c) \quad (48)$$

The Jacobi matrix  $\left\| \frac{\partial \vec{f}_c}{\partial \vec{r}} \right\|$  is then:

$$- \sum_p \frac{m_p}{|\vec{r}_c - \vec{r}_p|^3} \begin{bmatrix} 1 - \frac{3(x_c - x_p)(x_c - x_p)}{|\vec{r}_c - \vec{r}_p|^2} & -\frac{3(y_c - y_p)(x_c - x_p)}{|\vec{r}_c - \vec{r}_p|^2} & -\frac{3(z_c - z_p)(x_c - x_p)}{|\vec{r}_c - \vec{r}_p|^2} \\ -\frac{3(x_c - x_p)(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} & 1 - \frac{3(y_c - y_p)(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} & -\frac{3(z_c - z_p)(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} \\ -\frac{3(x_c - x_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & -\frac{3(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & 1 - \frac{3(z_c - z_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} \end{bmatrix} \quad (49)$$

$$\left\| \frac{\partial f_c^i}{\partial r_c^j} \right\| = - \sum_p \frac{m_p}{|\vec{r}_c - \vec{r}_p|^3} \left[ \delta_{ij} - \frac{3(r_c^i - r_p^i)(r_c^j - r_p^j)}{|\vec{r}_c - \vec{r}_p|^2} \right] \quad (50)$$

With  $\delta_{ij}$  the identity matrix with  $\delta_{ij} = 1$  if  $i = j$  where  $i, j$  runs spatial index from 1 to 3. For example,  $r^1 = x$ ,  $r^2 = y$ , and  $r^3 = z$  as usual sense. (We do not consider covariant form of this because we are not considering spacetime).

The Hessian matrix  $\left\| \frac{\partial \vec{f}_c}{\partial \vec{r}_c \partial \vec{r}_c} \right\|$  is then:

$$\left\| \frac{\partial^2 f_c^x}{\partial r_c^i \partial r_c^j} \right\| = - \sum_p \frac{3m_p}{|\vec{r}_c - \vec{r}_p|^5} \begin{bmatrix} \frac{5(x_c - x_p)^3}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) & \frac{5(x_c - x_p)^2(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) & \frac{5(x_c - x_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) \\ \frac{5(x_c - x_p)^2(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) & \frac{5(x_c - x_p)(y_c - y_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) & \frac{5(x_c - x_p)(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} \\ \frac{5(x_c - x_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) & \frac{5(x_c - x_p)(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & \frac{5(x_c - x_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) \end{bmatrix} \quad (51)$$

$$\left\| \frac{\partial^2 f_c^y}{\partial r_c^i \partial r_c^j} \right\| = - \sum_p \frac{3m_p}{|\vec{r}_c - \vec{r}_p|^5} \begin{bmatrix} \frac{5(x_c - x_p)^2(y_c - y_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) & \frac{5(x_c - x_p)(y_c - y_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) & \frac{5(x_c - x_p)(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} \\ \frac{5(x_c - x_p)(y_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) & \frac{5(y_c - y_p)^3}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) & \frac{5(y_c - y_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) \\ \frac{5(x_c - x_p)(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & \frac{5(y_c - y_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) & \frac{5(y_c - y_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) \end{bmatrix} \quad (52)$$

$$\left\| \frac{\partial f_c^z}{\partial r_c^i \partial r_c^j} \right\| = - \sum_p \frac{3m_p}{|\vec{r}_c - \vec{r}_p|^5} \begin{bmatrix} \frac{5(x_c - x_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) & \frac{5(x_c - x_p)(y_c - y_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & \frac{5(x_c - x_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) \\ \frac{5(x_c - x_p)(y_c - z_p)(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} & \frac{5(y_c - y_p)^2(z_c - z_p)}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) & \frac{5(y_c - y_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) \\ \frac{5(x_c - x_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(x_c - x_p) & \frac{5(y_c - y_p)(z_c - z_p)^2}{|\vec{r}_c - \vec{r}_p|^2} - 3(y_c - y_p) & \frac{5(z_c - z_p)^3}{|\vec{r}_c - \vec{r}_p|^2} - 3(z_c - z_p) \end{bmatrix} \quad (53)$$

$$\left\| \frac{\partial^2 f_c^i}{\partial r_c^j \partial r_c^k} \right\| = - \sum_p \frac{3m_p}{|\vec{r}_c - \vec{r}_p|^5} \left[ \frac{5(r_c^i - r_p^i)(r_c^j - r_p^j)(r_c^k - r_p^k)}{|\vec{r}_c - \vec{r}_p|^2} - \frac{3}{w} (\delta_{ij}(r_c^k - r_p^k) + \delta_{jk}(r_c^i - r_p^i) + \delta_{ik}(r_c^j - r_p^j)) \right] \quad (54)$$

where  $w = \delta_{ij} + \delta_{jk} + \delta_{jk} + \epsilon_{ijk}$  and  $\epsilon_{ijk}$  is 3D Levi-Civita symbol. We add Levi-Civita symbol to avoid zero in denominator Here again, latin indices  $i, j$ , and  $k$  indicates spatial components

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**Algorithm 2** Gravitation computation

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```
1: procedure TREE_TRAVERSAL_GRAV(branch sink)
2:   if sink.mass <  $M_{cellmax}$  then                                     ▷ Another choice criterion can be use
3:      $\vec{f}_c \leftarrow \vec{0}$ 
4:      $\frac{\partial \vec{f}_c}{\partial \vec{r}} \leftarrow \vec{0}$ 
5:     TREE_TRAVERSAL_C2C(sink, tree.root,  $\vec{f}_c$ ,  $\frac{\partial \vec{f}_c}{\partial \vec{r}}$ )                 ▷ Compute  $\vec{f}_c$  and  $\frac{\partial \vec{f}_c}{\partial \vec{r}}$  using MAC
6:     SINK_TRAVERSAL_C2P(sink,  $\vec{f}_c$ ,  $\frac{\partial \vec{f}_c}{\partial \vec{r}}$ )                             ▷ Expand to the particles below
7:   else
8:     for All children c of sink do
9:       TREE_TRAVERSAL_GRAV(c)
10:    end for
11:  end if
12: end procedure
13:
14: function MAC(branch sink, branch source, double macangle)
15:    $d_{max} \leftarrow source.radius \times 2$ 
16:    $dist \leftarrow distance(sink.position, source.position)$ 
17:   return  $d_{max}/dist < macangle$ 
18: end function
19:
20: procedure TREE_TRAVERSAL_C2C(branch sink, branch source, acceleration  $\vec{f}_c$ )
21:   if MAC(sink, source, macangle) then
22:      $\vec{f}_c \leftarrow \vec{f}_c + (-\frac{source.mass \times (sink.position - source.position)}{|sink.position - source.position|^3})$ 
23:      $\frac{\partial \vec{f}_c}{\partial \vec{r}} \leftarrow \dots$ 
24:   else
25:     if source.is_leaf() then
26:       for All particles p of source do
27:          $\vec{f}_c \leftarrow \vec{f}_c + (-\frac{p.mass \times (sink.position - p.position)}{|sink.position - p.position|^3})$ 
28:          $\frac{\partial \vec{f}_c}{\partial \vec{r}} \leftarrow \dots$ 
29:       end for
30:     else
31:       for All children c of source do
32:         TREE_TRAVERSAL_C2C(sink, c,  $\vec{f}_c$ )
33:       end for
34:     end if
35:   end if
36: end procedure
37:
38: procedure TREE_TRAVERSAL_C2P(branch current, acceleration  $\vec{f}_c$ )
39:   if current.is_leaf() then
40:     for All particle p of current do
41:        $p.grav \leftarrow \vec{f}_c + \frac{\delta \vec{f}_c}{\delta current.position} \cdot (p.position - current.position) + \dots$ 
42:     end for
43:   else
44:     for All children c of current do
45:       TREE_TRAVERSAL_C2P(c,  $\vec{f}_c$ ,  $\frac{\partial \vec{f}_c}{\partial \vec{r}_c}$ )
46:     end for
47:   end if
48: end procedure
```

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