#### Exercises for the lecture

# Fundamentals of Simulation Methods

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#### Exercise sheet 4

## A tree code for gravity

## 1. Tree code for gravity calculation

Tree algorithms can approximately calculate the forces in an N-body system by means of a hierarchical multipole expansion. This reduces the computational cost to something of order  $\mathcal{O}(N \ln N)$ . On the download page of the lecture, a simple skeleton tree code can be downloaded (a C-version as well as a Python-version are provided). Note that this implementation uses only monopole order and is not optimized for speed or memory consumption. You can use this template for this exercise, either directly or translated to another language, or if you wish, you may also write your own tree code from scratch using the template as an example (it's fun to do! ;-) ).

Consider N particles of total mass  $M_{\rm tot}=1$  placed randomly into a cubical box of unit side-length. For definiteness, we assume all particles have the same mass, and we adopt G=1. Assume that the gravitational potential of individual particles is Plummer-softened with a gravitational softening length  $\epsilon=0.001$ , i.e. we adopt the potential

$$\phi(\mathbf{r}) = -\frac{m}{(\mathbf{r}^2 + \epsilon^2)^{1/2}} \tag{1}$$

for a single particle of mass m.

Calculate the gravitational forces for all N particles with an oct-tree and determine the typical force accuracy and calculational time in comparison with direct summation, both as a function of N and of the opening angle parameter  $\theta^*$ . To this end, carry out the following steps:

- (a) The idea of this exercise is that you use one of the two templates provided (tree.c or tree.py & call\_tree.py). But if you like a challenge, you are encouraged to write your own version!
- (b) The provided code template is incomplete, so start by filling in the missing pieces. In particular, you need to add the computation of the centers of mass and total masses of tree nodes from their subnodes, as well as the partial force calculation from a tree node that is used in the tree walk. Also, please add a calculation of the *exact* forces by direct summation. When you're done, verify that the force approximation delivered by the tree code is roughly correct by adding suitable output to the code.
- (c) To allow a more quantitative analysis, add an automatic measurement of the average force error after all forces have been calculated by the tree and direct summation.

To this end, consider the relative force error

$$\eta = \frac{|\mathbf{a}_{\text{tree}} - \mathbf{a}_{\text{direct}}|}{|\mathbf{a}_{\text{direct}}|} \tag{2}$$

for each particle, and determine a simple arithmetic average  $\langle \eta \rangle$  of the mean relative force error<sup>1</sup>. Also, add diagnostic code that tells you the average number of particle-node interactions per particle, i.e. how many nodes the multipole expansion on average involves.

- (d) Make a little grid of calculations for N=5000, 10000, 20000 and 40000, and opening angles  $\theta^*=0.2$ , 0.4 and 0.8. In each case, measure the calculation time for the tree-based force calculation and for direction summation, as well as  $\langle \eta \rangle$  and the mean number of terms the tree code used per particle. Report the results in a table. **NOTE:** For direct summation with N=40000 particles the calculation could take very long; you can extrapolate from the smaller N cases since we know it scales as  $N^2$ .
- (e) Using the previous results, make a plot of the execution time of the force calculation with the tree as a function of N (for the  $\theta^* = 0.4$  case). Use logarithmic axis and fit a regression line (in the log) to the 4 data points you obtained. Also include the results for direct summation. Estimate the time needed for  $10^{10}$  particles for both methods by extrapolating your timing results.

<sup>&</sup>lt;sup>1</sup>Note that a more careful analysis of the errors should really consider the full distribution of the errors. It could well be that there are occasionally very large, catastrophic errors that go unnoticed in a simple average such as as  $\langle \eta \rangle$ . To diagnose this, one needs to look at the error distribution function.