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ABSTRACT
The performance and accuracy of a GRAPE-3 system for collisionless N-body simulations is discussed. After an initial description of the hardware configurations available to us at Marseille, and the usefulness of on-line analysis, we concentrate on the actual performance and accuracy of the direct summation and tree code software. For the former we discuss the sources of round-off errors. The standard Barnes-Hut tree code cannot be used as such on a GRAPE-3 system. Instead particles are divided into blocks and the tree traversal is performed for the whole block, instead of for each particle in the block separately. Then the forces are calculated by direct summation over the whole interaction list. The performance of the tree code depends on the number of particles in the block, the optimum number depending on the speed of the front end and the number of boards. We find that the code scales as $O(N)$ and explain this behaviour. The time per step decreases as the tolerance increases, but the dependence is much weaker than for the standard tree code. Finally we find that, contrary to what is expected for the standard version, the speed of our tree code increases with the clustering of the configuration. We discuss the effect of the front end and compare the performance of direct summation and tree code on GRAPE-3 with that of other software on general purpose computers.

The accuracy of both direct summation and the tree code is discussed as function of number of particles and softening. For this we consider the accuracy of the force calculation as well as the energy conservation during a simulation. Because of the increased role of the direct summation in the force calculation, our tree code is much more accurate than the standard one. Finally we follow the evolution of an isolated barred galaxy using different hardware and software in order to assess the reliability and reproducibility of our results. We find excellent agreement between the pattern speed of the bar in direct summation simulations run on the high precision GRAPE-4 machines and that in direct summation simulations run on our GRAPE-3 system. The agreement with the tree code is also very good provided the tolerance values are smaller than about 1.0.

We conclude that GRAPE-3 systems are well suited for collisionless simulations and in particular those of galaxies. This is due to their good accuracy and their high speed which allows the use of a large number of particles.

Key words: galaxies: structure – galaxies: kinematics and dynamics – methods: numerical.

1 INTRODUCTION

N-body simulations have come a long way since the pioneering work of Holmberg (1941). The ever-increasing computing power available has allowed them to become very useful tools for the understanding of the formation, dynamics and evolution of galaxies, and, particularly if used in concert with analytical work and confrontation to observations, can trigger important progress in the field. For this reason many different codes have been written to date, each aiming for a better accuracy and/or performance. Recent reviews of this quest have been given by Sellwood (1987) and Athanassoula (1993). The simplest approach, direct summation, is the one containing the least approximations. In this case the force on a given particle is obtained simply by adding the contributions from all other particles in the configuration. The big disadvantage of this method is that it is prohibitively expensive in CPU time, since the computational cost is pro-
portional to \( N(N - 1) \), where \( N \) the number of particles in the configuration. For this reason it could not be used up till now with a number of particles sufficient for most applications. This situation changed drastically with the advent of the GRAPE boards, which are special-purpose boards on which are calculated the gravitational forces between particles. These are linked to a front end machine, performing all the remaining tasks in the simulations. In this way very high performance can be achieved. A good description of the GRAPE project and the technical description of the boards can be found in a number of papers written by the team that conceived them (e.g. Sugimoto et al. 1990, Ebisuzaki et al. 1993, Makino et al. 1997).

In this paper we evaluate the performance and accuracy of the Marseille GRAPE-3 systems, and compare them with those of general purpose computers. The outline of the paper is as follows: We give the description of the configurations in section 2. We are using two types of software, direct summation (sec. 3.1) and the tree code (sec. 3.2). We discuss in detail the performance of the latter in section 4 and in general the timing of our GRAPE-3AF system in section 5. We next discuss the accuracy of the direct summation (sec. 6) and of the tree code (sec. 7). Section 8 is devoted to comparisons of results obtained with these two methods and of results obtained on GRAPE-4 hardware. The example chosen is the long term evolution of a barred galaxy.

2 DESCRIPTION OF THE MARSEILLE GRAPE SYSTEMS

Our group acquired a hand-wired GRAPE-3A card in March 1993. This consists of 4 LSI chips operating at 20 MHz and gives a peak speed equivalent of more than 2 Gflops. It is coupled via a Solfower SFVME-110 Sbus/VMEbus converter to a Sparc 10/412 workstation, which drives it. The configuration is illustrated in the left panel of Figure 1. This hardware is limited to 32 768 particles. It is, however, possible to use more particles than the hardware limit simply by dividing the total number of particles into packages of 32 768 or less, presenting one package at a time to the board, and then adding the contributions of all packages on the front end machine. Further information on the GRAPE-3A chips and the relevant software can be found in Ebisuzaki et al. (1993) and Makino & Funato (1993).

In August 1994 we acquired five GRAPE-3AF boards, each having 8 chips, giving us a peak speed equivalent of more than 20 Gflops. These are printed-circuit boards and occupy the five slots of a Solfower SFVME-110 Sbus/VMEbus converter, which links them to the host machine. As such we have used consecutively a Sparc 10/41, a Sparc 10/512, an Ultra 1/170 and, since August 1996, an Ultra 2/200 with two processors. The usefulness of the second processor will be discussed later in this section. The configuration is illustrated in the right panel of Figure 1. The GRAPE-3AF cards are hardware limited to 131 072 particles, yet it is possible, as for the GRAPE-3A boards, to use them for a larger number of particles with the help of appropriate software.

Every GRAPE-3A or GRAPE-3AF board has one memory unit. In case of multiboard systems, as for our GRAPE-3AF system, the host computer sees the system as one unit, since the memory units share the same address space. Thus when the host computer sends data to the memory, these are stored identically on the memory units of all boards. Different boards calculate forces on different particles, so that if a system has \( N_{\text{chips}} \) chips the forces will be calculated simultaneously for \( N_{\text{chips}} \) particles. Because of the nature of the N-body problem it is easy to use all chips in parallel, since the forces on different particles can be calculated independently.

Considerable gain in time can be achieved if the data transmission overlaps partly with the force calculation on the chips (Okumura et al. 1993). Thus when the data has been loaded to the first board this board can start calculating, without waiting for the data to be loaded on the second board etc. This amounts to a significant saving in the case of five boards, as in the Marseille system.

Since GRAPE-3 boards are meant to be used only for collisionless simulations they use low accuracy arithmetic. 14 bits are used to represent the masses, 20 bits for the positions and 56 bits for the forces. As will be argued below, the accuracy thus obtained is sufficient for most collisionless simulations. A more extensive discussion of the round-off errors is given in section 3.

The output from our simulations consists of masses, positions and velocities of all particles, as well as in some cases, potentials. For real*4 accuracy this means 32 bytes per particle and time step saved. To keep the necessary disc space to manageable limits we have recourse to on-line analysis, in which we anticipate which physical quantities need to be extracted from the data. Storing these can be done relatively frequently, so that the analysis can be done with adequate time resolution. Such analysis can include the amplitude and phase of several Fourier components of the density of individual galaxies, their energy and angular momentum, or that of their subcomponents, or whether individual galaxies in a cluster or group have merged or not. Furthermore bitmaps containing frames of the particle positions at a given time

![Figure 1. Schematic layout of our two GRAPE-3 systems, on the left GRAPE-3A and on the right GRAPE-3AF.](image-url)
step can be stored to construct short “movies”, which are helpful in revealing features of interest.

We found that a most suitable way to handle the load of the on-line analysis is to use a second processor on the same workstation. The first processor running the GRAPE boards then spawns at regular intervals a task starting analysis scripts. Certain tasks, such as calculating the amount of mass still bound to a companion or to a given galaxy in a group, involve potentials or forces and are calculated faster on a GRAPE board. In such cases we have recourse to the computer which runs the GRAPE-3A board.

3 SOFTWARE

3.1 The direct method

The direct method on GRAPE-3 is quite straightforward. The GRAPE hardware calculates the pairwise interactions and sums the forces on each particle from all other particles in the system. More detailed information, together with some examples, has been given by Makino & Funato (1993). Here we will discuss how the round-off errors are generated in GRAPE-3.

The round-off error in the GRAPE-3 hardware has two main origins (Makino, Ito & Ebisuzaki 1990). The first one is the round off error generated when the positions are converted from floating-point number format to fixed-point format. GRAPE-3 performs the subtraction \( x_i - x_j \) in 20-bit fixed-point format, and a resolution of 1/1024 is typically used, where the size of the system is taken to be of order unity. For the pairwise force, the relative round-off error due to this is \( O(10^{-3}/r_{ij}) \). In other words, the error is larger for nearby pairs than for far-away ones.

The second source of round-off error is the calculation of the force from the relative position vector. This part of the calculation is implemented in logarithmic format with an accuracy of around 1%, and has a r.m.s. error of 2% (Okumura et al. 1993). This error does not depend on \( r_{ij} \).

In conventional general-purpose computers additional round-off error is generated in the summation, since one typically adds small numbers (the pairwise forces) to a large number (the calculated total force). However, since GRAPE-3 uses a 56-bit fixed point format for the summation no additional round-off errors are generated here.

3.2 The tree code

The version of tree code we use is essentially the same as the vectorisation scheme described by Barnes (1990). The main difference from the standard algorithm (Barnes & Hut 1986) is that the particles are first divided into blocks and then the tree traversal is performed for a block of particles, instead of for each particle in the system. Then GRAPE hardware is used to calculate the forces from the nodes in the interaction list created by this tree traversal to all particles in the block. Note that this procedure is essentially the same as the construction of the “local essential tree” in the distributed-memory parallel version of the tree code developed by the Caltech group (Salmon & Warren 1994). However, the way we use the obtained data is quite different. In the case of the parallel tree code, the force on each particle is still calculated by traversing the tree, while in the case of the GRAPE tree code, we calculate the force by direct summation over the whole interaction list.

The speed of the tree traversal, which is performed on the front end, depends on the number of particles in the block. Generally speaking, if we make the number of particles in a block larger, we can reduce the amount of work of the front end, but we increase the cost of the calculation on GRAPE, since the average length of the interaction list becomes longer (Makino 1991). Thus, the number of particles in a block should be chosen so that the total calculation cost is minimum. This optimal number depends mainly on the relative speed of GRAPE and the front end, but also, though to a smaller extent, on the number of particles and their distribution in the system.

It is intuitively expected that the block size should also influence the accuracy, since the interaction list of a given block will contain at least as many members as the interaction list of any particle in the list and probably quite a bit more. This was tested by Barnes (1990), for values of \( n_{\text{crit}} \) less than 256, and will be further examined in section 7.1 for values of \( n_{\text{crit}} \) of a few thousand, which will be shown in section 4.1 to be optimum for our GRAPE configuration.

A further difference with the Barnes-Hut (1986) standard algorithm is that instead of the conventional multipole acceptability criterion (MAC) our code uses the minimum distance MAC (Salmon & Warren 1994). In this criterion instead of the distance between a body and the center of mass of a cell one uses the minimum distance from the body to any point in the cell. This has the useful aspect that it is completely independent of the contents of the cell and of the center of the mass of the bodies in it. A further advantage of this criterion according to Salmon & Warren (1994) is that it does not admit the rare, yet not impossible, unbound error they found for the conventional MAC.

Since GRAPE-3 can only calculate the forces and potentials between pairs of particles, the tree code has only the monopole term and neglects quadrupoles and higher order terms. However, as we shall see in sections 7 and 8, this does not prohibit it from reaching a satisfactory accuracy level, as one can, without much additional cost, decrease the tolerance or increase the number of particles.

4 PERFORMANCE OF THE TREE CODE

4.1 Dependence on the number of particles in a block

In this section we examine more closely how the number of particles in a block influences the performance of the tree code. In order to divide the particles into blocks we descend the tree and regard a cell as being a group if the number of particles it contains is less than a given number, \( n_{\text{crit}} \). While its parent cell contains more particles than \( n_{\text{crit}} \). Thus the number of particles in a block can vary, but must necessarily be smaller than \( n_{\text{crit}} \). Makino (1991) showed that for a Plummer sphere* and for values of \( n_{\text{crit}} \) roughly in the range

* The density of the Plummer sphere is...
10 < n_{\text{crit}} < N/100 \text{ (where } N \text{ the total number of particles)} \text{ the number of particles per block is roughly } n_{\text{crit}}/4.

As already mentioned in the preceding section, the performance of the tree code will depend on the number of particles in the block and therefore on n_{\text{crit}}. We also expect it to depend on the total number of particles and the tolerance. In order to have some quantitative estimates of how these three parameters influence the performance we have measured the time necessary for one time step for several values of n_{\text{crit}} as well as of the tolerance and of the total number of particles. These times were obtained by averaging over 20 time steps in order to decrease the effect of statistical fluctuations. Plotting these times as a function of n_{\text{crit}} we find a minimum, as expected, but it is very shallow and thus ill defined. For N = 500 000 we find that the optimum range for n_{\text{crit}} is between 5 000 and 25 000, and this range shrinks towards smaller values as N decreases, so that e.g. for N = 150 000 it is between 3 000 and 15 000. For a large number of particles the value of the tolerance does not seem to influence the optimum value of n_{\text{crit}}. For N around 100 000, however, larger values of the tolerance correspond to smaller optimum values of n_{\text{crit}}. The fact that the minimum is so shallow leaves a lot of freedom in the choice of the optimum n_{\text{crit}}. The trade-off in this case is between memory requirements and accuracy, since for large n_{\text{crit}} the results are more accurate, but necessitate more memory, than for low n_{\text{crit}}.

Seen the above, we can conclude that a reasonable compromise for our GRAPE-3AF system is for n_{\text{crit}} around say 7 000 or 8 000.

\[ \rho(r) = \frac{3M}{4\pi b^3} \left(1 + \frac{r^2}{b^2}\right)^{-5/2} \]

where M and b are its mass and scale length.

### 4.2 Dependence on the total number of particles

Figure 2 shows the CPU time per time step as a function of the total number of particles in the system, N, and for various values of the tolerance. As expected, this time increases with N, but what is interesting to note is that this increase is linear. For the standard tree code Barnes and Hut (1986) argued that it should have an Nlog(N) dependence, while Hernquist (1987) showed that for his vectorised implementation the CPU time was proportional to aNlog(N) + bN for a tolerance \( \theta \geq 0.4 \). In order to understand the time dependence of our tree code we use the analysis of Makino (1991) which shows that the time for the force calculation should be proportional to \( a(n_{\text{crit}}, \theta)N + 100N\log(N\theta^3/23)/\theta^3 \), where \( a(n_{\text{crit}}, \theta) \) a function of n_{\text{crit}} and \( \theta \), given by eq. (7) of Makino (1991).

For the values of \( n_{\text{crit}} \) of 0.9, 1.0, and 1.1 used in our cases the first terms is much bigger than the second one, so that the time dependence should be proportional to N, in good agreement with the results of Fig. 2.

### 4.3 Dependence on the tolerance

We have examined the values \( \theta = 0.5, 0.6, 0.7, 0.8, 0.9, 1.0 \) and 1.1. As expected, the time per step decreases as the tolerance increases, and that for all values of \( n_{\text{crit}} \) and of N. This decrease, however, is relatively less important than that found for a standard tree code (Hernquist 1987). Furthermore, it is not equally important for all values of the tolerance, and in particular is relatively small for \( \theta > 0.8 \). This is clear from Fig. 2, where the curves corresponding to \( \theta \geq 0.8 \) cluster together.

### 4.4 Dependence on the clustering of the points

All the tests given above, as well as most given so far in the literature, concern a Plummer sphere distribution. Yet conventional wisdom says that the performance of the tree code should depend heavily on how the points are distributed. In order to test this we measured the CPU per time step in configurations of variable clumpiness.

For this we considered 10 points (which can be considered as galaxy centers) randomly distributed in a given volume \( D^3 \) and around each a number of points distributed with a radial density profile \( \exp(-\alpha r) \) (which can be considered as representing a “galaxy”). For large values of \( \alpha D \) the configurations will of course be more concentrated around the 10 centers, than for small values of \( \alpha D \). We find that the CPU time necessary for one time step decreases with increasing clustering, but that the effect is relatively small. Thus for \( N = 25 000, \theta = 0.5, n_{\text{crit}} = 8 000, D =20 \) and \( \alpha = 0.5, 1., 2. \) and 5. we have respectively 0.97, 0.90, 0.82 and 0.79 seconds. For \( N = 500 000 \) and the same values for the other parameters we have 31.23, 31.09, 30.76 and 29.78 seconds. This trend goes against the conventional wisdom that more clustered configurations should necessitate longer CPU times, but the differences can be easily understood in terms of the differences between the standard Barnes & Hut tree code and our GRAPE version of it. Indeed in the
Barnes & Hut tree code a more centrally concentrated configuration means that the tree descent has to go deeper and is therefore more time-consuming. This, however, is not the case for our version of the tree code since for the particles in the same block we use direct summation and therefore it makes no difference whether they are centrally concentrated or not. On the other hand for a system with a large value of $\alpha_D$ “galaxies” will be very centrally concentrated and therefore will need few subdivisions for the tree when seen from a point in another “galaxy”. Thus more clustered configurations should necessitate somewhat less CPU time than less clustered ones, as was indeed seen in the numerical examples.

5 TIMING OF GRAPE-3AF

5.1 The effect of the front end

In order to see how much of a difference a faster front end can make to the speed of our simulations we have repeated the beginning of two simulations with the front ends at our disposal and compared the running time. The first simulation is a direct summation case with 65 400 particles. It is run $k_164$ in the notation of Athanassoula, Makino & Bosma (1997) and represents a compact group of five identical Plummer galaxies distributed according to a King $\Psi = 1$ law. 87.5% of the mass of the system is in a common halo which follows a similar mass distribution and encompasses the whole group. We find that the Sparc 10/512, the Ultra 1 and the Ultra 2 take 6.26, 5.90 and 5.89 seconds per time step respectively. The differences are small, due to the fact that most of the work in the direct summation case is done by the GRAPE system.

Our second example is the evolution of a barred galaxy with a live halo, represented by 120 000 particles in total. In this case we follow the evolution with a tree code with an opening angle of 0.6 and no quadrupole terms. We find that the Sparc 10/512, the Ultra 1 and the Ultra 2 take 17.26, 8.2 and 7.9 seconds per time step respectively. In this case the difference between the performances of the three machines is more important than in the previous example. A more meaningful comparison, however, would involve only the time spent on the front end, so that we have to subtract from the total time that used for the force calculations on GRAPE and the communication time between the host and the boards. About 12 Mbytes of data have to be transferred between the host and GRAPE at every time step. Since the effective transfer rate of the transfer box is about 3-4 Mb/sec, the data transfer would take about 3-4 seconds. Roughly extrapolating from Makino (1991) we find the length of the interaction list to be around 3 000, which, for the 120 000 particles in our system, gives $3.6 \times 10^8$ interactions. Taking into account that our 5 board GRAPE-3AF system has a speed of $8 \times 10^8$ interactions/second, we find that the time spent for the calculations is less than a second. Thus the total time spent for communications and calculations on GRAPE is of the order of 4 seconds and most of it can be accounted for by the communications. This gives us that the times spent on the front end are 3.9, 4.2 and 12.9 seconds for our Ultra 2, Ultra 1 and the Sparc 10/512 respectively.

![Figure 3. CPU time necessary for one time step, as a function of the number of particles in the system. Results are shown for direct summation on our GRAPE-3AF system (dashed lines and circles with crosses), the tree code on our GRAPE-3AF system with a tolerance of 0.7 (solid line and open circles), a tree code on the front end and with the same value of the tolerance (dot-dashed lines and circles with a dot) and a three dimensional cartesian grid code with a 129x129x129 grid (dotted lines with lozenges) and with a 257x257x257 grid (dashed-dotted-dotted lines and diamonds).](image)

The ratio of the times on the Ultras is in good agreement with the ratio of their clock speeds. For a comparison with the Sparc 10/512 we have to use specfp values, so that the comparison is not as straightforward. Nevertheless the ratios agree to better than a factor of two.

5.2 GRAPE-3AF compared to other potential solvers

Fig. 3 compares the CPU times necessary for one time step as a function of the number of particles in the system and different codes. The GRAPE timings were obtained using our GRAPE-3AF system with direct summation and tree code respectively. For the tree code on the front end we used the version available in NEMO, which is due to Josh Barnes. For the cartesian grid code we used Jerry Selwood’s code, which uses Richard James’s potential solver.

The break-even point between direct summation and tree code on our GRAPE-3AF system is around 25 000 particles. For a higher number of particles the tree code is faster.
This number of course depends on the ratio of the CPU performance of GRAPE and of the front end and would thus be different for another front end or a different number of boards.

Comparison of the NEMO scalar tree code and the GRAPE tree code shows that the GRAPE tree code is about 30 times faster for a small number of particles and about 40 times faster for larger numbers. We have not searched for the tree code which would run fastest on our front end. The version we are using, however, has the advantage of not being optimised for a vector machine, which would unnecessarily hamper its performance on a scalar machine. Also the version of the tree code which is running on our GRAPE-3AF system is a rather straightforward implementation of a previously existing version and it should be possible to achieve considerable gains in performance by rewriting the code according to the needs of the GRAPE boards. Such a work is in progress.

For the cartesian grid code we have considered two different resolutions, a high resolution grid of 257x257x257 and a low resolution one of 129x129x129. We have not been able to extend the tests for the high resolution grid code beyond 100 000 particles, because of memory limitations, since grid codes require considerable memory allocations. Nevertheless we can deduce that the high resolution grid code runs considerably slower than our GRAPE tree code for all number of particles tested and even slower than our direct summation GRAPE for less than the order of 40 000 particles. On the other hand the low resolution grid code runs faster than the direct summation for more than 100 000 particles and faster than the tree code for more than 300 000 particles.

It is thus clear from the above diagram that our GRAPE codes, both direct summation and tree code, give very high performances, and can be used with a very large number of particles, the last statement being particularly true for the tree code.

6 Accuracy of Direct Summation on GRAPE-3AF

6.1 Accuracy of the force calculation

Merritt (1996) and Athanassoula et al. (1997, in preparation) discussed the value of the softening ($\epsilon$) which gives the best approximation of the force due to a given density distribution. For this they use the quantity

$$MISE = <\int \rho(x)|F - F_{\text{true}}(x)|^2 \, dx>$$

and a density distribution corresponding to a Plummer sphere of unit mass and unit scale length. We repeated the exercise using direct summation on GRAPE-3AF. Since a Plummer distribution is spherically symmetric the above simplifies to a one-dimensional integration along a radius. For this we use the alternative extended Simpson’s rule (Press et al. 1988), which has an accuracy of $O(N^{-4})$, and 100 points along the line of integration. The upper limit of the integration was taken to be $L = 20b$, where $b = 1$ is the scale length of the Plummer sphere. This radius contains more than 99% of the mass of the Plummer model, while there the density has fallen at that point to roughly $3 \times 10^{-7}$ of its central value. The number of realisations was taken to be $6 \times 10^6/N$. More details on this calculation are given in Athanassoula et al. (1997, in preparation). The results are shown in Fig. 4, and coincide within the mean errors with the 64 bits direct summation results. This could, at first sight, sound at odds with the fact that GRAPE-3AF is a low precision machine. Nevertheless one should keep in mind that the error in GRAPE-3AF comes from round-off and thus can be considered random. To illustrate this we calculated the force at the center of a Plummer sphere. The position of a minimum error along a line for a given $N$ is marked by an X, and the corresponding $\epsilon$ value is the optimum softening $\epsilon_{\text{opt}}$ for this number of particles.

![Figure 4. MISE as a function of the softening $\epsilon$ for a Plummer sphere. From top to bottom the curves correspond to $N = 30, 100, 300, 1000, 3000, 10000, 30000, 300000$, where $N$ is the number of particles in the realisation of a Plummer sphere. The position of a minimum error along a line for a given $N$ is marked by an X, and the corresponding $\epsilon$ value is the optimum softening $\epsilon_{\text{opt}}$ for this number of particles.](image)
minimum $MISE$ value for a given number of particles $N$ simply by fitting a second order polynomial to the three points with the lowest $MISE$ values. We thus found the optimum softening length, $\epsilon_{opt}$, and the corresponding $MISE$ value as a function of the number of particles $N$ and we display them in Fig. 6. Both can be well represented by power laws:

$$\epsilon_{opt} = 0.98N^{-0.26}$$ (1)

and

$$MISE = 0.22N^{-0.68}$$ (2)

There is, however, an indication that $\log(MISE)$ is not a linear function of $\log N$, but that a second order polynomial would be a better fit. Thus for small numbers of particles the error increases with decreasing $N$ less fast than for large number of particles. For that reason the exponent of eq. (2) will depend somewhat on the range of $N$ considered.

Our results agree nicely with those of Merritt (1996). In particular we find the same exponent for the dependence of $\epsilon_{opt}$ on $N$, while we find that the dependence of $MISE$ on $N$ is somewhat steeper, which, taking into account the effect discussed in the above paragraph, can be understood, since our results extend to higher numbers of particles. This point will be taken up again in section 7.1.

### 6.2 Simulations of a Plummer sphere

In the above we discussed the effect of the softening on the calculated value of the force. However the accurate representation of the force is only one of the aspects to be taken into account in numerical simulations. Furthermore the largest contribution to the $MISE$ or $MASE$ comes from the immediate neighbourhood of each particle or point, while classical theory of two-body relaxation tells us that it is the contribution of distant particles that dominates the relaxation effect. For this reason we evolved a 100 000 particles Plummer sphere using direct summation on the GRAPE-3AF boards and five different values of the softening, namely 0.01, 0.03, 0.05, 0.1 and 0.2 and checked energy conservation. Our Plummer sphere is in virial units, i.e. $G = 1$, $M = 1$ and $4E = -1$, where $G$ the gravitational constant, $M$ the total mass of the Plummer sphere and $E$ its total energy (kinetic plus potential). This gives a scale length of $b = 3\pi/16$. The softening value for which the force is optimally described for 100 000 particles is $0.047 \ast 3\pi/16 = 0.028$, i.e. within the range of values tried and very near the second value 0.03. The upper panel of Fig. 7 shows the relative energy difference as a function of time for the five simulations evolved with a time step of $\delta t = 1/64 = 0.015625$. In order to lessen the noise and bring out trends more clearly we applied nine point sliding means to the data. We see that the case with the highest softening, where the force calculation contains a considerable bias, starts out of equilibrium and within the first few time steps readjusts. Thus the energy evolves fast to a value much different from the initial one. The three intermediate cases have good energy conservation, the best one being the case with a softening of 0.03, which is the value nearest to the one predicted by the minimum of the $MISE$ as a function of $\epsilon$ curve (cf. Fig. 4). For the case with the smallest softening the energy is badly conserved, showing a steady increase with time. This, however, is not due to the value of the softening, but to the combination of the value of the softening and that of the time step. Indeed we used a ratio of $\epsilon/\delta t = 0.64$ while for our Plummer sphere $<u^2>^{1/2} = 0.7$. We thus repeated the simulations for half and a quarter of the time step, i.e. $\delta t = 1/128$ and 1/256. The results are plotted in the lower panels of Fig. 7, and show that for a sufficiently small time step the energy can be well conserved.

Since both a good conservation of the energy and an accurate representation of the force are important for high quality numerical simulations the choice of the appropriate
$\epsilon$ is a complex matter. The softening that gives the most accurate representation of the force can be found by calculations such as those in the previous subsection. On the other hand a good energy conservation depends on both the softening and the time step. However a small value of the softening imposes the choice of a small time step, which in turn imposes higher CPU time requirements.

7 ACCURACY OF THE TREE CODE ON GRAPE-3AF

7.1 Accuracy of the force calculation

In order to quantify the accuracy of the tree code and compare it to that of direct summation we have used the quantity $MASE$, introduced by Merritt (1996, cf. also Athanassoula et al. 1997).

$$MASE = \frac{1}{N} \sum_{i=1}^{N} |F_i - F_{true}(x_i)|^2 >$$

This is similar to $MISE$, used in section 6.1, but the force is now calculated on all particles in the configuration, rather than at some points on a line. Thus it is much more time-consuming to calculate than $MISE$, but this does not pose in our case any problem, since the tree code on GRAPE is very fast. For the same reason, and in order for the range of values of $N$ to be comparable to what we can use for simulations with the corresponding code on GRAPE, we have examined a different range of values of $N$ than what was used for direct summation. Namely we have not considered values less than 30 000, which would, anyway, be meaningless with the adopted values of $n_{crit}$, while we extended the upper end for $N$ to 1 000 000, a value which as we saw in section 5.2, can be used without problem on our GRAPE configuration.

The results obtained with $n_{crit} = 4000$ and $\theta = 0.5$ are given in Figures 8 and 9. The first thing to note by comparing Figures 4 and 8 is that the accuracy of the tree code is comparable to that of direct summation. This can be understood since the main source of error in $MISE$ or $MASE$ comes from the nearby particles and, since these are treated by direct summation in both cases, we get comparable values for the two methods.

As shown in Fig. 9 both the optimum softening length $\epsilon_{opt}$ and the corresponding $MASE$ value, $MASE_{opt}$, can be represented as power laws of the number of particles $N$

$$\epsilon_{opt} = 0.63N^{-0.22}$$

(3)

and

$$MASE_{opt} = 0.38N^{-0.73}$$

(4)

These numbers are considerably different from those found in section 6.1. The difference, however, is not due to a difference between the two codes but to a difference between the range of number of particles considered in the two cases. This is made clear in Table 1, where we give the values of the exponent for $\epsilon_{opt}$ (column 3) and $MASE_{opt}$ (column 4), together with the corresponding range of particle numbers

$N$ (column 2). The first line repeats the values from Merritt (1996) and the second one those of section 6.1. For the third and fourth line we have considered separately two ranges of particle numbers, both for direct summation. The fifth line gives our values for the tree code. Finally the sixth line gives the asymptotic values obtained for $N \rightarrow \infty$, as calculated in the Appendix. We note that indeed as the range of particle numbers becomes higher the exponents approach their asymptotic limit.

We also calculated $MASE$ values for different values of the tolerance and of $n_{crit}$. The results came out as expected, i.e. the force calculations were more precise for smaller tolerances or larger $n_{crit}$, but the effects were small. The fact that, contrary to the standard tree code, the effect of the tolerance on the accuracy is small, can be easily explained by the fact that the largest contribution to $MISE$ or $MASE$ comes from relatively nearby particles, for which the force in our tree code is anyway calculated by direct summation. This argues that, in order to get a higher accuracy, it is preferable to increase the number of particles rather than to decrease the tolerance.

7.2 Simulations of a Plummer sphere

We repeated the five simulations of the previous section, using this time the tree code instead of direct summation. The results were quite satisfactory, although the tree code simulations conserved energy somewhat less well than the direct summation ones, as could be expected. The run of the energy with time can be mentally decomposed into some global trend on which is added some noise. The average relative value of this noise was of the order of 4 parts in 10$^5$ and did not seem to depend on the value of the softening.

The global trend gave an energy conservation of 4 parts in 10$^5$ for $\epsilon = 0.01$ and a time span $\Delta t = 100$, and less than 2 parts in 10$^5$ for the remaining values of the softening.

8 LONG-TERM EVOLUTION OF A BARRED GALAXY

In this section we discuss the evolution of the same initial conditions with different methods, in order to be able to assess the effect of the code on the results. For this we use initial conditions corresponding to a bar-unstable disc galaxy, where the halo is described by 120 000 particles and the disc by 60 000, and evolve it using direct summation on
Figure 7. Relative energy difference, \((E(t) - E(0))/E(0)\), as a function of time for five simulations of the evolution of a Plummer sphere realization with 100 000 particles. Each evolution corresponds to a different softening, 0.01 (solid line), 0.03 (dashed line), 0.05 (dash-dotted line), 0.1 (dotted line), and 0.2 (dash-dot-dot-dot line). A horizontal dashed line at zero relative energy difference has been plotted to guide the eye. The upper panel corresponds to a time step of 1/64, the middle one to 1/128, and the lower one to 1/256.

Figure 8. \(MASE\) as a function of the softening \(\epsilon\) for a Plummer sphere and a tree code with \(\theta = 0.5\) and \(n_{\text{crit}} = 4\ 000\). From top to bottom the curves correspond to \(N = 30\ 000, 100\ 000, 300\ 000,\) and 1 000 000, where \(N\) is the number of particles in the realization of the Plummer sphere. The position of a minimum error along a line for a given \(N\) is marked by an X, and the corresponding \(\epsilon\) value is the optimum softening \(\epsilon_{\text{opt}}\) for this number of particles.

Figure 9. Optimum softening length, \(\epsilon_{\text{opt}}\) as a function of number of particles (squares and solid line; scale on the ordinate on the right) and corresponding \(MASE\) values (crosses and dashed line; scale on the ordinate on the left).

GRAPE-4 (for a description of this high accuracy machine, see Makino et al. 1997), direct summation on GRAPE-3AF, and tree code on GRAPE-3AF with 5 different values of the tolerance. Quantities that allow us to make quantitative comparisons between the different results are the pattern speed and the amplitude of the bar. For this we simply measure the phase and amplitude of the \(m = 2\) component as a function of radius on-line in all three cases with the help of the same software. This allows us to calculate the pattern speed from the smoothed time derivative of the phase, averaged over the radii where the bar is best defined. The results are displayed in Fig. 10. The observed difference is due to the exchange of angular momentum between the bar on the one hand and the disc and halo on the other, and its interpretation, as well as the discussion of its importance, have been the subject of many papers (e.g. Weinberg 1985, Hernquist & Weinberg 1992, Littler & Carlberg 1991, Athanassoula 1996, Sellwood & Debattista 1996). Here we will only be interested in how well the results of the various codes agree with each other. In the upper left panel we compare the results of GRAPE-3 with those of GRAPE-4, and find excellent agreement. A comparison between the GRAPE-3 results and those with a tree code and a tolerance of 0.5 (not shown here) are also very satisfactory. In the upper right panel we compare tree code results with tolerances of 0.5, 0.7, 1.0, 1.2 and 1.5. The values obtained with the two biggest values of the tolerance are considerably smaller than those obtained with the other values. The lower left panel compares the results obtained in all cases with a tree code with a tolerance of 0.7 and different number of particles, 180 000, 90 000 and 45 000 respectively. We note that the values of the pattern speed obtained for 45 000 particles are considerably lower than the others, while the other two are relatively close, arguing that they are converging and therefore that 180 000 particles are sufficient for such a simulation.

Finally the lower right panel compares the results in the case when all the particles in the simulation are not of the same mass. In one case we have considered particles in the
Figure 10. Pattern speed of the bar as a function of time. The upper left panel compares results obtained with GRAPE-4 (solid line) and GRAPE-3AF (dashed line). The upper right panel compares results obtained with the tree code and tolerances of 0.5 (solid line), 0.7 (dashed line), 1.0 (dash-dotted line), 1.2 (dotted line) and 1.5 (dash-dot-dot-dotted line). The lower left panel compares simulations with the tree code and a tolerance of 0.7 for 180,000 particles (solid line), 90,000 particles (dashed line) and 45,000 particles (dash-dotted line). The lower right panel again refers to simulations with a tree code and a tolerance of 0.7. The solid line corresponds to a simulation with equal mass particles in the disc and in the halo, the dash-dotted line to a simulation where the disc particles are twice as massive as those of the halo and the dashed line to a simulation where the mass of the particles in the halo is four times as big as that of the particles in the disc.

Disc which are twice as massive as those in the halo and in the other particles in the halo which are four times as massive as those in the disc. This last case gives results which are considerably lower than those of the other two. The “trick” of using more massive and therefore fewer particles for the halo, in order to have more particles in the disc, has been often used in numerical simulations, but the above results show that such simulations should be interpreted with caution. Of course one can argue that, since we do not know what the halos are made of, we do not know what value to use for the ratio of the mass of the halo particles to that of the disc particles. Nevertheless it should be kept in mind that the adopted value could influence considerably the results. For example in merging simulations heavier halo particles could sink faster to the center than lighter ones, and thus lead to a more concentrated final halo. A measure of the bar amplitude is given in Fig. 11 as a function of time. The layout of the different panels is the same as that of Fig. 10. We again note that the GRAPE-3 and GRAPE-4 results agree very well. Also the bar amplitude is smaller in the cases with higher tolerance values, as could be expected since higher tolerance gives more smoothing. The same thing is seen also for simulations with a lower number of particles. Finally, as for the previous figure, the mass of the particles in the halo influences considerably the results.

9 SUMMARY

In this paper we present the Marseille GRAPE-3 systems and discuss their possibilities and limitations. At present we have a system with 5 GRAPE-3AF boards linked
Figure 11. Measure of the $m = 2$ component of the density as a function of time. The layout and the various line styles are the same as for the previous figure.

In order to keep the storage requirements to a reasonable level it is mandatory to use on-line analysis. The complication here, compared to simulations run on workstations, is that GRAPE is very much faster than the front end, so it is unreasonable to have it wait while the slow front end does the analysis. The problem is solved if the workstation has a second processor which is assigned to do the on-line analysis. In cases where this analysis involves heavy calculations that can be done on a GRAPE, we have found it useful to use our slower GRAPE-3A system for that task. The on-line analysis we perform includes the production of short movies, giving visual information on the simulation, as well as the calculation of several quantities for the different components in isolated galaxies or the different galaxies in pairs, groups or clusters. We give increasingly more importance to the on-line analysis and its proper planning can take as much time as the preparation of the initial conditions.

On our GRAPE systems we run two kinds of simulation software, direct summation and the tree code, and we give here a description of each. In the case of the tree code particles are divided in blocks with a common interaction list and then direct summation is used over this list. We analyse the performance of the tree code and how this depends on various parameters such as the number of particles in a block, the total number of particles, the tolerance and the clustering of the points. We thus find the optimum number of particles per block for our system to be of the order of 7000 to 8000. Contrary to conventional wisdom, we find that less clustered configurations take longer CPU times and this can be explained by the differences between the conventional Barnes-Hut tree code and our version which uses direct summation between particles in the same block.

A faster front end brings little improvement to the performance of the direct summation code, since most of the time is taken by calculations on the boards. The opposite is true for the tree code, where a faster front end can make...
considerable difference, as we show by analysing the relative times on the front end and on the boards.

We then compare the performances of our GRAPE system with that of a tree code and a cartesian grid code run on the front end. As expected, we find that the GRAPE tree code is a very significant improvement compared to the front end version, but even direct summation goes much faster than the front end tree code, at least for a number of particles up to a million or more. Both our codes are also considerably faster than a high resolution (257x257x257) grid code. Only the low resolution (129x129x129) cartesian grid code beats our tree code for more than 300 000 particles. Furthermore, a considerable improvement in the performance of our tree code can be expected if the code is rewritten according to the specifications of the GRAPE boards.

A powerful potential solver can be useless if it does not have sufficient accuracy. For this reason we analyse extensively the accuracy of both direct summation and tree code. For the accuracy of the force calculation we use the concepts of MISE and MASE introduced by Merritt (1996), which allow a clear estimation of how accurately the force is calculated. We find that the forces are as accurate as when full accuracy is used on a front end. The reason is that errors in the force calculations on GRAPE are due to round-off and are thus purely random. Thus they cancel out when we sum the force contributions from a large number of particles. This means that results obtained with the GRAPE boards will have the full accuracy of direct summation with 32 or 64 bit precision and argues that, in order to increase the accuracy of the force calculation, one should increase the number of particles in the realisation rather than consider a more accurate potential solver, if, as is the case for GRAPE, the errors are not systematic.

Again using the MASE values we find that the accuracy of the GRAPE tree code is comparable to that of direct summation and that can be explained by the fact that contributions of nearby particles are calculated in both cases in the same way. As expected the force calculations are more accurate for smaller tolerances or larger crit, but the effects are small. In order to increase the accuracy it is thus more efficient to increase the number of particles than to decrease the tolerance.

We also performed a number of simulations of the evolution of a Plummer sphere in order to test the energy conservation, although the latter does not depend only on the calculations on the GRAPE boards but also on the time integration scheme and time step used. We find very good energy conservation and discuss the influence of the softening on the time step that should be used.

As a final test of the adequacy of GRAPE-3 boards to stellar dynamical simulations we evolve the same initial conditions using different hardware and software. These include the 64-bit precision direct summation on GRAPE-4 boards and direct summation and tree code on GRAPE-3 boards. The simulation is the long term evolution of a barred galaxy and we find that the pattern speeds as calculated by direct summation on GRAPE-4 and on GRAPE-3 show excellent agreement. The agreement is also very good with the tree code with small opening angles. Using large opening angles is not as satisfactory.

One can thus conclude from all the above that GRAPE-3 boards are well suited for simulations of galaxies or galaxy systems, both because they have the necessary accuracy and because their high speed allows the use of a large number of particles.

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APPENDIX A1: ASYMPTOTIC BEHAVIOUR OF THE MEAN INTEGRATED SQUARE ERROR

The asymptotic behaviour of the MISE or MASE can be understood with a very simple argument as below. Let us first discuss the “bias” part (Merritt 1996) of the error, which is the error due to the finite softening in the limit of \( N \to \infty \), and then the effect of finite \( N \), or the “variance” part in the terms used by Merritt (1996).

In the following, we consider the softening in terms of the softening kernel function \( W(\frac{r}{\epsilon}) \). The exact force on a particle at position \( r \) is given by

\[
F_{\text{exact}} = \int dr' F_{pp}(r, r') \rho(r').
\]  
(A1)

where \( F_{pp} = G \frac{r - r'}{|r - r'|^3} \). The softened force is calculated as

\[
F_{\text{soft}} = \int dr' \rho(r') \int dx W(\frac{x}{\epsilon}) F_{pp}(r, r' + x). \]  
(A2)

Since \( F_{pp} \) actually depends only on the relative position, we can rewrite the above equation as

\[
F_{\text{soft}} = \int dx W(\frac{x}{\epsilon}) F_{\text{exact}}. \]  
(A3)

By changing the order of the space integration, we then find

\[
F_{\text{soft}} = \int dx W(\frac{x}{\epsilon}) F_{\text{exact}}. \]  
(A4)

Equation (A4) has the same form as that used in all definitions in smoothed particle hydrodynamics (SPH, e.g. Monaghan 1992), where \( W \) is the smoothing, or interpolating kernel. Following the argument used in SPH (e.g. Hernquist & Katz 1989), we can show that

\[
F_{\text{soft}} = F_{\text{exact}} + \mathcal{O}(\epsilon^2), \]  
(A5)

so that for the squared error we have

\[
|F_{\text{soft}} - F_{\text{exact}}|^2 \propto \epsilon^4. \]  
(A6)

For the “variance” part of a non-singular distribution of particles we can write

\[
|F_{\text{nbody}} - F_{\text{soft}}|^2 \propto (N \epsilon)^{-1}. \]  
(A7)

Equation (A7) can be understood as follows: For a given softening, if one changes \( N \), the squared error should decrease as \( 1/N \), since we can consider the N-body force as a Monte-Carlo integration of the softened potential field. The argument for the dependence on the softening is equally simple. The random error in the force is dominated by the variation of the forces from particles with the distance of the order of the softening parameter. The number of particles in that region is proportional to \( \epsilon^3 \). On the other hand, the typical force from one of these particles is proportional to \( \epsilon^{-2} \). Thus, the squared error is proportional to \( \epsilon^{-1} \). Similar results, both for the “variance” and the “bias” have been found by Merritt & Tremblay (1994).

To derive the “optimum” value of \( \epsilon \), which will minimise the sum of the bias and the variance, we write

\[
MISE = c_1 \epsilon^4 + \frac{c_2}{N \epsilon}, \]  
(A8)

where \( c_1 \) and \( c_2 \) are constants. Thus, the optimal \( \epsilon \) and \( MISE \) are given by

\[
\epsilon_{\text{opt}} \propto N^{-0.2}, \]  
(A9)

\[
MISE_{\text{opt}} \propto N^{-0.8}. \]  
(A10)

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