Time stepping N-body simulations

Thomas Quinn, Neal Katz\(^1\), Joachim Stadel, and George Lake

*Astronomy, Box 351580, University of Washington, Seattle, WA 98195*

**ABSTRACT**

Leapfrog integration has been the method of choice in N-body simulations owing to its low computational cost for a symplectic integrator with second order accuracy. We introduce a new leapfrog integrator that allows for variable timesteps for *each* particle in large N-body simulations. Tests with single particles in fixed potentials show that it behaves as a symplectic integrator. We then examine the results of both standard leapfrog and our temporally adaptive leapfrog on full N-body integrations of clusters and large scale structure establishing accuracy criteria for both methods. The adaptive method shows significant speed-ups over single step integrations—but the integrator no longer appears to be symplectic or, in the case of large scale structure simulations, accurate. This loss of accuracy appears to be caused by the way that the timestep is chosen, not by the integrator itself. We present a related integration technique that does retain sufficient accuracy. Although it is not symplectic, it is apparently better than previous implementations and is our current integrator of choice for large astrophysical simulations. We also note that the standard leapfrog difference equations used in cosmological N-body integrations in comoving coordinates are not symplectic. We derive an implementation of leapfrog that is in comoving canonical coordinates to correct for this deficiency.

*Subject headings: Methods: numerical*

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\(^1\)Current address: Department of Physics and Astronomy 517 Lederle Graduate Research Tower University of Massachusetts Amherst, MA 01003-4525
1. Introduction

Over the past decade, spatially adaptive methods have been developed to calculate gravitational forces in N-body simulations. These include tree codes (Appel, 1985; Barnes and Hut, 1986), fast multipole methods (Greengard, 1987), and adaptive grids (Couchman, 1991). As the number of particles in an N-body simulation grows, so do the density contrasts. Hierarchical methods can follow extremely large dynamic ranges in densities at modest additional cost per force evaluation. However, large ranges in densities also imply a large range in time scales ($\propto 1/\sqrt{\text{density}}$). If we take the final state of a simulation and weight the computational work done on particles not uniformly but inversely with their natural timesteps, we find a potential gain of $\sim 50$. Temporal adaptivity is one of the last algorithmic areas where we can target an order of magnitude improvement. Hence, we seek a hierarchical integrator, i.e., a method such that particles are on adjustable individual timesteps.

The most commonly used time integration scheme for N-body simulations is the leapfrog method. Leapfrog has several advantages over other methods. 1) For second order accuracy only one force evaluation and one copy of the physical state of the system is required. This is particularly beneficial for N-body simulations where the cost of a force evaluation is very expensive. 2) The force field in an N-body simulation is not very smooth, so higher order does not necessarily mean higher accuracy. 3) It is a symplectic integrator, i.e., it preserves properties specific to Hamiltonian systems. See Channell and Scovel (1990) for a review of symplectic integrators. Gravitational N-body systems are Hamiltonian, and therefore they should benefit from the use of an integrator that conserves phase space volume and has no spurious dissipation. This could be especially important in self-gravitating systems where a dissipation time scale that is linked to the dynamical time can lead to a runaway to spuriously large densities. Leapfrog is a second order symplectic integrator requiring only one costly force evaluation per timestep and only one copy of the physical state of the system. These properties are so desirable that we concentrate on making an adaptive leapfrog.

Hierarchical leapfrogs have been used before (Porter 1985; Ewell 1988; Hernquist and Katz 1989), but they were not symplectic (§4). There exist symplectic integrators with individual but fixed timesteps for either particles or modes (Saha & Tremaine 1994; Skeel & Biesiadecki 1994; Lee, Duncan & Levison 1997). Hut, Makino & McMillan (1995) proposed an iterative scheme for choosing timesteps for a single particle in a rapidly varying potential, but each iteration involves a force evaluation that is prohibitively expensive for large N simulations (§2).

Section 2 describes the theory behind a hierarchical integrator. Section 3 describes tests of this integrator on a single particle in a potential, and Section 4 will present tests of the integrator on full N-body systems. We discuss the implications of these results in Section 5.

2. Symplectic Integrators

A symplectic integrator is an exact solution to a discrete Hamiltonian system that is close to the continuum Hamiltonian of interest. Therefore, it preserves all the Poincaré invariants and places stringent conditions on the global geometry of the dynamics. An obvious example is total energy conservation in a system with a time-independent Hamiltonian or the conservation of angular momentum in axisymmetric systems (Zhang & Skeel 1995). A symplectic integrator will exactly conserve the energy in the discrete Hamiltonian that is an approximation to the true energy of the system. This approximate energy oscillates about the true energy without any numerical dissipation.

The difference between the discrete and continuum Hamiltonians can be viewed as a small
perturbation given by the truncation error of the integrator. The error is a Hamiltonian! If the error Hamiltonian is a sufficiently small perturbation, then the KAM theorem (Arnold 1978) guarantees that the invariant curves destroyed are a set of finite measure. In other words, almost all orbits that are stable in the real system will continue to be stable in the numerical system. An illustration of these advantages is shown in Figure 1. Here the radial velocity, \( v_r \), is plotted against the radius, \( r \), for an ellipticity, \( e = 0.5 \) Kepler orbit using a leapfrog integrator and using a fourth order Runge-Kutta integrator. In each integration, approximately 24 steps were taken per orbit, and the integrations ran for 16 orbits. Note how the leapfrog integrator oscillates about the true solution but always remains on a one dimensional surface. This indicates that it is indeed conserving an energy-like quantity, \( i.e. \) having the orbit constrained to a one dimensional surface shows the existence of an isolating integral of motion. On the other hand, the Runge-Kutta orbit slowly becomes more circular. The poor performance of the Runge-Kutta integrator is remarkable given that it is a fourth order integrator and uses four times as many force evaluations as the leapfrog integrator. Also note the large wiggles in the leapfrog integration at apoapse. These are indicative of the proximity of resonant islands that would lead to an instability for larger timesteps.

The leapfrog integrator can be written as

\[
\begin{align*}
    r_{n+1/2} &= r_n + \frac{\tau}{2} v_n, \\
    v_{n+1} &= v_n + \tau a(r_{n+1/2}), \\
    r_{n+1} &= r_{n+1/2} + \frac{\tau}{2} v_{n+1},
\end{align*}
\]

where \( r \) is the position vector of a particle, \( v \) is the velocity, \( a \) is the acceleration, and \( \tau \) is the timestep. When several of these steps are put together, the two position updates can be combined to a single update \( r_{n-1/2} \) to \( r_{n+1/2} \), and the resulting alternation between updating \( r \) and \( v \) gives leapfrog its name. The symplectic nature of leapfrog can be seen by noting that the position update is equivalent to evolving the system exactly under the Hamiltonian \( H_D = \frac{1}{2} v^2 \), and the velocity update is equivalent to evolving the system under the Hamiltonian \( H_K = V(r) \), where \( V(r) \) is the potential generating the accelerations. We will call the operator that evolves the system under \( H_D \) the “drift” operator, \( D \), and the operator that evolves the system under \( H_K \) the “kick” operator, \( K \).

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It can be shown that (see Saha and Tremaine, 1992 for details) the combination of operators \( D(\tau/2)K(\tau)D(\tau/2) \) evolves the system under a Hamiltonian

\[
H_N = H_D + H_K + H_{err} = \frac{1}{2} v^2 + V(r) + H_{err},
\]

where \( H_{err} \) is of order \( \tau^2 \). The existence of this surrogate Hamiltonian ensures that the leapfrog

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2 If one expresses Hamilton’s equations as \( \dot{z} = \{z, H\} \), where \( z \) are the phase space coordinates and \( \{ , \} \) are Poisson brackets, then formally \( D(\tau)z = \exp(\tau\{z, H_D\}) \), and \( K(\tau)z = \exp(\tau\{z, H_K\}) \).
is symplectic and second order. Since the Hamiltonian is symmetric with respect to \( H_D \) and \( H_K \), the combination of operators \( K(\tau/2)D(\tau)K(\tau/2) \) is also a symplectic second order integrator. Explicitly, this is

\[
\begin{align*}
  v_{n+1/2} &= v_n + \frac{\tau}{2} a(r_n), \\
  r_{n+1} &= r_n + \tau v_{n+1/2}, \\
  v_{n+1} &= v_{n+1/2} + \frac{\tau}{2} a(r_{n+1}).
\end{align*}
\]

Higher order symplectic integrators can be constructed from combinations of leapfrog steps (Yoshida 1990). Each \( N \)th order leapfrog integrator requires \( N - 1 \) force evaluations and only one copy of the physical state of the system.

Unfortunately, constructing a variable stepsize method by choosing a new timestep after each leapfrog step gives very disappointing results (Calvo and Sanz-Serna, 1993). Some simple schemes can be shown to have no rigorous stability criterion for any step size (Skeel 1993). Several explanations have been given for this behavior. The simplest is to note that a variable step integrator is evolving a dynamic system with state variables \((r, v, \tau)\). The projection onto the phase space coordinates \((r, v)\) cannot be described by a Hamiltonian. Skeel and Gear (1992) consider a one step symplectic operator of the form

\[
(r_{n+1}, v_{n+1}) = F(r_n, v_n, \tau(r_n, v_n)),
\]

and show that if \( F \) is symplectic for a constant step size, it will not in general be symplectic for variable step size. Another way to see this problem is to notice that the time reversibility has been broken: if we step forward in time and then step backward, we do not end up at the same point because of the change in timestep. Since symplectic implies time reversible, such an operator is not symplectic. Integrators that are time reversible are referred to as reflexive by Kahan (1993; see also Kahan and Li 1997) who argues that reflexivity is the key to the robust properties of “updating formulae” rather than symplecticity.

A strategy to make leapfrog reflexive are clear from a cursory look at its evolution operators. The operation \( D(\tau/2)K(\tau)D(\tau/2) \) is reversible if we use an “select” operator to choose the timestep, \( S \), such that the time reversibility is retained. Hut, Makino, and McMillan (1995) achieve this by using an implicit definition of the timestep:

\[
\tau = \frac{1}{2} \left[ \tau(r_n, v_n) + \tau(r_{n+1}, v_{n+1}) \right].
\]

So, the beginning and end of the step are required to agree on the timestep. One can solve for \( \tau \) iteratively, and very good results are obtained even with only one or two iterations. (This result is also clear from Kahan 1993). This iterative scheme poses several problems when applied to a large \( N \)-body simulation. It requires backing up timesteps, throwing away expensive force calculations, using auxiliary storage and must specify a method for synchronizing the particles for mutual force evaluations.

However, we can also see an alternative means to restore reflexivity. Let us choose a select operator \( S \) that commutes with \( K \), so that \( DSKD \) is equivalent to \( DKSD \). Since \( K \) only changes the velocities and not the positions, an \( S \) operator that depends entirely on positions satisfies the commutation requirement. As we shall show below, this is not strictly time reversible. However, since the operators read the same forward and backward, we will refer to \( DSKD \) as a “palindromic” integrator. Synchronization can be maintained by only choosing timesteps that are a power-of-two subdivision of the largest timestep, \( \tau_s \). That is,

\[
\tau_i = \frac{\tau_s}{2^n_i},
\]

where \( \tau_i \) is the timestep of a given particle, and \( n_i \) is an integer (See Hernquist and Katz, 1989). Combining this synchronization procedure with the \( DSKD \) evolution operation we have the following (recursive) algorithm for a timestep. 1) Drift the particles forward \( \tau_i/2 \). 2) Apply the
select operator. If it accepts this step, then finish the step with $K(\tau_i)$ and $D(\tau_i/2)$. Otherwise, drift the particles back $\tau_i/2$ and take two timesteps using $\tau_i/2$. The algorithm can be expressed recursively in the following pseudo-code:

```c
Timestep(\tau) {
  Drift(\tau);
  Select(\tau);
  if \tau = \tau_{\text{min}}
    { Kick(\tau);
      Drift(\tau);
    }
  else
    { Drift(-\tau);
      Timestep(\tau/2);
      Kick(\tau);
      Timestep(\tau/2);
    }
}
```

Here, $\text{Drift}(\tau)$ applies $D(\tau)$ to all particles, $\text{Kick}(\tau)$ applies $K(\tau)$ to particles with timestep $\tau$ and $\text{Select()}$ decides to which timestep the particles belong, with the side effect of finding the minimum timestep, $\tau_{\text{min}}$. We will refer to the traditional method of choosing a timestep at the beginning of the integration step as $\text{SDKD}$.

As an example of how this procedure works, consider Figure 2. In this diagram we consider the case where there are particles on three separate timesteps. The arcs represent drifting the particles, and the vertical slashes represent either an select or kick of particles at that timestep. Starting at point 0, all particles are drifted to point 1 and evaluated for whether they are on the largest timestep. Then all particles are drifted to point 2 where particles on the middle timestep are kicked, then to point 4 where particles on the smallest timestep are kicked, then to point 1 where particles on the largest timestep are kicked. Then all particles are drifted to point 5 where again all particles not on the largest timestep are evaluated for whether they are on the middle timestep. Then the appropriate particles are kicked at points 6, 5, and 7. Finally all particles are drifted to point 8 and a single large timestep is complete.

There are several things to note about the above algorithm. One is that it is not unique. It uses a “top down” approach by trying the largest timestep and reducing it until it is deemed appropriate. A “bottom up” scheme can be implemented where a smallest timestep is tried first. Secondly, this algorithm is not exactly time reversible. Scenarios can easily be constructed where a forward step followed by a backward step will not come back to the initial conditions. An example is shown in Figure 3. The circle delineates the boundary between region 1, where one timestep is needed, and region 2, where a timestep of 1/2 the timestep in region 1 is needed. If we start a particle in region 2, it is drifted forward, found to still be in region 2, so two timesteps are taken. If we start the particle in
region 1 and go backward, it is possible that the drift leaves the particle still in region 1, so only one timestep is taken going backward, and we do not arrive at the point from which we started. Finally, since several selects may be done for every kick, the efficiency of this algorithm depends upon Select being able to quickly decide an appropriate timestep given the particle positions. A suitable criterion that depends only on the positions is one based upon the local dynamical time\( t_d = 1/\sqrt{G\rho} \). With this criterion, Select will pick the largest timestep \( \tau \) such that

\[
\tau < \frac{\eta}{\sqrt{G\rho}}.
\]

where \( \eta \) is a constant to be determined based on stability and accuracy requirements. With this criterion, region 2 in Figure 3 is a high density region while region 1 has low density. Similar problems to this loss of reflexivity can occur in the iterative schemes (Hut, Makino, & McMillan 1995; Kahan 1993) if the starting guess of an iteration is preconditioned by the history of the particle. That is, the loss of reflexivity depends as much on the choice of a “top down” approach as on the Select operator.

3. Single Particle Tests

Our first test is the motion of a test particle in the one dimensional effective potential of the Kepler problem. If the integrator performs well, these tests will guide our choice of \( \eta \) for later applications. The timestep criterion in equation 1 is not straight-forward to implement in this case, since the density is zero anywhere outside the central source. Therefore, in this test we determine the timestep by the enclosed density, \( \rho_e = 3M/4\pi r^3 \) where \( M \) is the mass of the central object.

Figure 4 shows the relative change in total energy as function of time for several integrators. The orbit has eccentricity \( e = 0.5 \), and is evolved for about 100 periods. Results for two values of \( \eta \) for the DSKD method are plotted as well as a result for SDKD. A fixed step integration was also run, and its energy nearly coincides with the
\( \eta = .03 \) DSKD method. From the energy plot, it appears that the DSKD method with \( \eta = 0.03 \) is indeed symplectic. In contrast, the SDKD method with the same timestep criterion experiences a secular drift in the energy. The savings in force evaluations is also significant. A fixed timestep integration with similar energy errors required 50000 force evaluations to accomplish 100 orbits while the adaptive timestep integration required less than 16000, a savings of over a factor of 3 for this moderate eccentricity orbit. Also, it can be seen from the energy plot that \( \eta \approx 0.03 \) is needed for the adaptive algorithm to be stable. This is because a stable orbit, having one integral of motion for each degree of freedom, should be quasiperiodic, which the \( \eta = 0.03 \) line appears to be. For \( \eta \) larger than this, the method does not appear to be symplectic, but the changes in energy appear to be due to instability, (i.e. integrals being destroyed) rather than a secular buildup of truncation error.

A more realistic test of the integrator is the motion of a single particle in the potential of an isothermal sphere. As well as having a density defined everywhere, allowing the use of the local density in the timestep criterion, the potential is similar to that of globular clusters, galaxies and clusters of galaxies in our large N-body simulations. Figure 5 shows the energy conservation for the SDKD and DSKD methods for a particle in the potential of a singular isothermal sphere. The ratio of apocenter to pericenter is 3.2:1. The density used for choosing the timestep is the local density of the isothermal sphere. Again the SDKD shows a secular drift in the energy, while the DSKD has the characteristics of a symplectic method. Also note that in the case of an isothermal sphere, the method is stable for \( \eta = 0.1 \).

\footnote{Remember that if there is one integral for each degree of freedom, the Hamiltonian can be expressed as \( H = H(J) \), and any function of the phase space coordinates, \( f(J, \theta) \) can be expressed as \( f(t) = f(J, \omega t) \) where \( J \) and \( \omega = \partial H/\partial J \) are constants.}

Fig. 5.— Relative change in total energy is plotted as a function of time in units of the orbital period for several integrators in an isothermal potential. Curves are plotted for DSKD with \( \eta = 0.1 \) and \( \eta = 0.03 \), SDKD with \( \eta = 0.03 \).

4. N-Body Tests

For our purposes, the real proof of a given integration method is how well it performs in an actual N-body code. In this section we test the above methods in two complementary systems. The first is a King model—a model which, in the absence of collisional relaxation, should remain static. The second is a cosmological simulation of a universe dominated by cold dark matter (CDM). This case is very dynamic, with structure evolving on all scales. All the simulations below were performed with PKDGRAV (Dikaiakos & Stadel, 1996; Stadel & Quinn, in preparation), a parallel code that uses a binary tree with a Barnes and Hut (1986) style opening criterion to make the gravity calculation order \( N \log(N) \). To mitigate the effect of force errors, we use an opening criterion \( \theta = 0.55 \), and accelerations from cells are expanded to hexadecapole
order in all the following simulations.

Before we test the multistep model in a full N-body simulation, let us explore the more basic question of what single fixed timestep is appropriate for an astronomically interesting N-body model. For our tests we will use a $W = 9$ King model realized with 100,000 particles evolved for 100 central dynamical times. This is representative of a galaxy halo over its lifetime in a typical cosmological N-body simulation.

Energy conservation is a typical measure of the quality of a simulation, but more appropriate measures should involve the convergence of scientifically interesting quantities. For the King model tests we will use the radial density profile which should remain constant in a collisionless simulation. In Figure 6 we compare the density profile of the initial King model with that at the end of a simulation with various timestep sizes evolved over 100 central crossing times. All models were integrated with fixed step leapfrog using timesteps of either 1.0, 0.3, 0.1, or 0.03 times the central dynamical time. From the figure, we see that $\tau < 0.1t_d$ is needed to maintain the central density of the King model. Figure 7 shows the evolution of the total energy for these same simulations. Note that an energy conservation of better than 3% is needed to preserve the density profile of this King model.

In order to use a density criterion to determine timesteps in a general N-body simulation, it is necessary to have a method for calculating the local density for every particle in an arbitrary distribution. We use a Smooth Particle Hydrodynamics estimate of the density by smoothing over a fixed number of nearest neighbors with a cubic spline kernel. The smoothing length, $h$, is adaptive and is set so that there are exactly...
Fig. 8.— The spherically averaged density is plotted against radius as in Figure 6. The integrations were done using a multistep leapfrog with a density timestep criterion. All integrations except the one indicated used a DSKD style timestepping.

64 particles within $2h$. The kernel is made symmetric using the “gather-scatter” algorithm as described in Hernquist and Katz (1989).

We show the radial density profiles for adaptive timestep integrations in Figure 8. From the figure, we see that, as in the single particle case, $\eta \leq 0.03$ is needed to integrate this model with reasonable accuracy. The DSKD integration with $\eta = 0.03$ needed a factor of 6 less force evaluations than the fixed step integration with $\tau/t_d = 0.03$, or a factor of 2 less than the fixed step, $\tau/t_d = 0.1$ integration. As shown in Figure 8, the energy conservation of the adaptive scheme is comparable to the fixed step integrator. Also shown in the figures are results for the SDKD integrator. It appears that at the level of accuracy needed to integrate this King model for 100 dynamical times, the secular drifts introduced by the SDKD integrator do not significantly affect the integration.

Before testing timestepping in a cosmological simulation we first note that the standard leapfrog difference equations used in cosmological simulations with comoving coordinates are not done in canonical coordinates and are apparently not symplectic. However, as shown in the appendix, a symplectic integrator can easily be derived, for which the “drift” and “kick” operators are

$$D(\tau) \equiv \mathbf{r}'_{t+\tau} = \mathbf{r}'_t + \mathbf{p}'_t \int_t^{t+\tau} \frac{dt}{a^2} \quad \quad K(\tau) \equiv \mathbf{p}'_{t+\tau} = \mathbf{p}'_t - \nabla\phi' \int_t^{t+\tau} \frac{dt}{a},$$

where the variables are as defined in the appendix. We use this integrator in the tests that follow. This integrator has another advantage over the standard leapfrog difference equations in that it can be used to implement both the $DKD$ and $KDK$ form of leapfrog. The standard equations can only be derived for $DKD$. The $KDK$
form of leapfrog can have advantages over $DKD$ (see the discussion of costs in §5).

Energy conservation may be an even poorer criterion for determining the accuracy of a cosmological integration. This owes to the possibility of making small errors in calculating the energy of the uniform background that can dominate the total error budget and mask any problems in the structure of interest. Nonetheless, it pays to examine the evolution of the energy. It was transients in the energy at the beginning of the cosmological simulations that made us suspect that the integration was not in canonical coordinates, leading to the results in Appendix A. But, these large transients in energy had little effect on the three dimensional structure, underscoring our point that it may be a poor diagnostic. Small changes in the way that two simulations handle global energy may be an interesting clue to other aspects of the integrator.

Since there is no analytic model for the full three dimensional formation of non-linear cosmological structure, we have to use another simulation to determine the ground truth. Therefore, we test the goodness of a given timesteping algorithm by examining its convergence to a simulation with a large number of timesteps for all particles. This convergence will be assessed based on properties of the groups that are formed and the internal structure of the largest group.

The fiducial simulation is of a standard CDM dominated universe in a periodic box of size $22.2$ Mpc ($h = 0.5$) on a side realized on a grid of $32^3$ particles. This implies a particle mass of $2.3 \times 10^{10}$ M$_\odot$. A spline softening is used with a softening length of 20 kpc. The initial conditions were imposed on the particles at a redshift of 49 using the Zel’dovich approximation. The small size of this box means it does not accurately represent large scale structure, but it does guarantee significant non-linear evolution and a stringent test for the time integrator.

The first convergence test is the density profile at the end of the simulation of the largest halo as identified using a friends-of-friends grouping algorithm. The object in question has a total mass of $1.23 \times 10^{14}$ M$_\odot$ within the virial radius and a maximum circular velocity of 639 kms$^{-1}$. This test looks at how the timestep algorithm affects the structure of well resolved objects in the simulation. Since structure forms hierarchically, the objects in the largest clusters will have experienced a complicated history with several dynamical times in previous generations of structure. Any effect that accumulates with dynamical time is most likely to betray itself in the center of the richest cluster.

Comparisons of the density profile of this object in various integrations are shown in Figure 10. The fixed step run used 4000 steps in a Hubble time. The $DSKD$ and $SDKD$ runs used $\eta = 0.03$. As might be expected, the timestep criterion used for the King model simulation performs well in accurately modeling the density profile of this cluster-size object. Looser criteria ($\eta \geq 0.1$) fail to capture the central density.
Fig. 11.— The cumulative mass fraction in groups of a given particle number or higher is plotted.

of the object, just as in the static King model.

Our second criterion is the cumulative mass fraction in groups, also as identified by the friends-of-friends algorithm with a linking length of 0.26 times the mean interparticle separation. This linking length corresponds to an enclosed density equal to the virial density assuming an isothermal sphere. Here the motivation is to see the effect of the timestep algorithm on the smallest objects we can resolve. The cumulative mass fraction of halos is plotted in Figure 11.

For the smallest objects, the density criterion does not do well. There is a significant decrease in the number of objects formed with less than one hundred particles in the simulation using the density timestep criterion. It may not be surprising that particles residing in halos with comparable or fewer particles than that used by the smoothing kernel are unable to accurately estimate their local density. On the other hand, a simulation using a (non-time symmetric) multistepping algorithm based on the acceleration of the particles reproduces the numbers of objects seen in the fixed step simulation all the way down to 8 particles. (The discrepancy seen at about 500 particles per group is due to a merger occurring at the end of the simulation.)

5. Discussion

First, we examine our results in terms of the number of timesteps needed to accomplish a given simulation with reasonable accuracy. We will determine the number of fixed steps before looking at the gain from hierarchical timesteping. For an equilibrium model where we wish to maintain the overall density profile, the answer is straightforward: the timestep should be set to \( \Delta t = \frac{\eta \sqrt{G\rho_{\text{max}}}}{v_c \sqrt{4\pi/3}} \), where \( \rho_{\text{max}} \) is the maximum density. For a cosmological simulation, more assumptions need to be made. Consider a simulation in which the largest halo has a circular velocity, \( v_c \), that is roughly constant with radius down to some core radius, \( r_0 \). If we wish to accurately model this halo, then our timestep criterion must be

\[
\Delta t = \frac{\eta}{\sqrt{G\rho_0}} = \frac{v_c \sqrt{4\pi/3}}{r_0},
\]

where \( \rho_0 \) is the (roughly constant) central density, and we have assumed spherical symmetry. Now consider a simulation with a halo with \( v_c = 1000 \, \text{kms}^{-1} \) where our goal is to resolve its structure within 10 kpc. In units of the Hubble time, \( 1/H_0 \), we have

\[
H_0 \Delta t \approx 1 \times 10^{-3} \eta \left( \frac{r_0}{10 \, \text{kpc}} \right) \left( \frac{v_c}{1000 \, \text{kms}^{-1}} \right)^{-1}.
\]

For \( \eta = 0.03 \), this gives 33,000 steps per Hubble time. Previous estimates of the timestep conclude that 6000(10 kpc/\( \epsilon \)) timesteps are needed per Hubble time, where \( \epsilon \) is the gravitational softening length (Lake et al. 1995). This is consistent with our estimate if we assume that the gravitational softening is 1/5 the core radius we wish to resolve.

To evaluate the performance of a multistep integrator, we must determine the “fixed costs” re-
quired for any level of the timestep hierarchy as well as the costs in determining the timestep level assigned to each particle. The total cost of force evaluations is then added and amortized over the longest timestep. For example, a tree code requires that the tree is built for all the particles regardless of the number of particles requiring force evaluations. Indeed, for a standard particle mesh code, the forces at every grid point are either calculated or not. The only part of the force evaluation that is not part of the fixed cost would be the trivial interpolation of the force onto particles. So, essentially all the costs are fixed and no gain is possible.

First, we consider the costs in the SDKD scheme with our tree code, PKDGRAV, and assume that the cost of actually moving the particles and evaluating the timestep is negligible. A single base step with different timesteps separated by a factor of two, will have a cost of

\[ (2^r - 1)C_t + C_f \sum_{i=1}^{r} 2^{i-1}N_i, \]

where \( C_t \) is the cost to build the tree, \( C_f \) is the cost of calculating the force on a single particle, and \( N_i \) is the number of particles on a timestep \( i \) of size \( 2^{1-i} \) times the base step. For a fixed timestep simulation with all particles at the smallest timestep, the cost would be

\[ 2^{r-1}(C_t + NC_f). \]

Comparing these two costs, it is easy to see that if \( C_t \) dominates the computation, then in the limit of large \( r \), the SDKD scheme would a factor of two more expensive than the single step calculation. On the other hand, if \( C_t \) is negligible, then the maximum speedup would be a factor of \( 2^{r-1} \) if almost all the particles were on the largest timestep. For a given ratio, \( f \), between \( C_t \) and the cost of evaluating all the forces, \( NC_f \), there is a maximum speedup of \((f+1)/2f\) for large \( r \), and almost all of the particles on the largest timestep.

Note that for an SKDK scheme this maximum speedup would be a factor of two larger since gravity evaluations for larger timesteps are synchronized with those of smaller timesteps, and the same tree can be used for both. For PKDGRAV on the King model with 100,000 particles, the ratio \( f \) is about 0.024, giving a maximum speedup of about 22.

For the symplectic scheme with a density criterion, the cost of evaluating the timestep is non-negligible, and we must account for it. It also has a fixed cost part for building the tree, \( C_{st} \), and a per particle cost, \( C_{sf} \). In terms of these costs the total cost of a base step is now

\[ (2^r-1)(C_t+C_{st})+C_f \sum_{i=1}^{r} 2^{i-1}N_i+C_{sf} \sum_{i=1}^{r} (2^i-1)N_i, \]

If the tree build costs and per particle costs for selecting the timestep and evaluating the forces are similar, then the symplectic scheme would be a factor of two more expensive than the non-symplectic scheme. For PKDGRAV running on a 100,000 particle King model, the per particle cost of the density criterion is a factor of 19 smaller than the per particle cost of a force evaluation, and the tree build for a density tree is 50% faster than for a gravity tree. The maximum speedup assuming an optimum particle distribution is then about 13.

For realistic particle distributions, the speedup can be much smaller, especially since a few particles (0.1% in the 100,000 particle King model) end up on timesteps smaller than the single stepping timestep. The speedup of the multistep run over the single step run as calculated from the above formulae is about a factor of 4. Note that we have neglected some costs such as particle pushing and the inefficiency of calculating forces for a few particles on modern pipelined processors. The actual factor in wall clock time is 2.7.

For non-equilibrium situations such as cosmological simulations, a multistep scheme offers additional speedups since it can adapt to the changing time scales in the simulation. An example of the timestep evolution is shown in Figure...
Fig. 12.— The cumulative fraction of particles on a given timestep or larger is plotted as a fraction of the age of the Universe in a cosmological simulation. The lowest line is for the longest timestep taken, and each upper line is for a factor of two smaller timestep.

Here, the fraction of particles at a particular timestep or larger is plotted as a function of time. Note how almost all particles start out on a very small timestep at the beginning of the simulation when the Universe is very dense, and generally migrate to larger timesteps as the simulation progresses. One could make adjustments for this trend in a single step simulation by transforming to an appropriate variable, e.g. expansion factor instead of time (as in Efstathiou et al. 1985), but as can be seen from the figure, a simple monotonic transformation would not capture all the complexities of the changing time scales. Furthermore, one would need different transformations for different cosmological models. Finally, to keep the integrations in canonical coordinates would require that the gravitational softening length is fixed in coordinates that were different from either comoving or physical (see Appendix A).

From our pragmatic standpoint as users of N-body simulations to model complex phenomena, the gain from symplectic integrators is their ability to tolerate greater truncation errors in numerical simulations of Hamiltonian systems. This, in turn, allows for longer timesteps and shortens the computer time needed to complete the simulation. Similarly, individual timesteps in a particle simulation allows one to concentrate the computational effort on those particles with the shortest dynamical times that require it. Either of these is beneficial only if their advantages outweigh their cost, whether in computational effort or algorithmic complexity.

For a system like the Solar System that is simulated for billions of dynamical times, a symplectic integrator has obvious advantages. Any dissipation introduced by truncation error has dire consequences that are easily observable—a planet may spiral into the Sun. The alternatives to symplectic integrators are either very high order integrators or extremely short timesteps, or both. Even a computationally expensive symplectic integrator can be an overall winner against these alternatives. (Wisdom and Holman, 1991; Saha and Tremaine, 1992.)

For other systems, such as galaxies or large scale structure, the situation is not so clear. These systems are simulated for at most a few hundred dynamical times, and drifts in conserved quantities due to truncation error will not be as noticeable. In these marginal cases, a symplectic integrator will only be beneficial if it is simple to implement and computationally cheap. In this paper we focus on a criterion based on the local density for this very reason: the algorithm for estimating the local density is very fast compared to calculating gravity. Unfortunately, this criterion is not ideal, particularly for cosmological simulations.

Is there a better criterion? If we use the kick-drift-kick form of leapfrog, one could imagine a timestep criterion that depends only on the ve-
locity, and therefore commutes with the drift operator. An example would be $\tau < \eta \epsilon/v$, where $\epsilon$ is the force softening. Three objections can immediately be raised against such a criterion. First, it is not Galilean invariant. A particle moving with respect to the coordinate system of the simulation will be on a different timestep than one that is at rest irrespective of the forces acting on the particle. Second, if the acceleration of a particle is nearly opposite to its velocity at the beginning of a timestep, then the timestep can be chosen such that the velocity is nearly zero at the middle of the timestep where the Select decision is made. Such particles will thereby be always put on a very long timestep. Third, the choice of $\epsilon$ to set the scale assumes that the dynamical times are set by two-body encounters. For a particle orbiting in a cluster, it is the cluster density rather than two-body encounters that set the dynamical time. On the other hand, in a cluster with significant substructure the timestep should be set by the encounters with this substructure. In this case, $\epsilon/v$ might be an accurate estimate of the timestep.

An obvious criterion is the acceleration, for example $\tau < \eta \sqrt{\epsilon/a}$. This commutes with the Kick operator because Kick changes only the velocities, and the acceleration is only a function of the positions. So one could easily use it as the select operator in DSKD just like we use the density criterion. However, the acceleration is just the thing that we are trying to minimize calculating. Using it as a timestep criterion will cancel the computational savings we are trying to achieve with multistepping. For an SDKD method, this is not a problem since one could use the last calculation of the accelerations that were used to advance the velocities. This luxury is not available for the DSKD method, and one would have to evaluate the acceleration of a particle many times during the Select phase. One could also imagine criteria based on higher order terms, such as the magnitude of the divergence of the acceleration, or the time derivative of acceleration. However, these tend to be even more expensive to calculate than the acceleration.

Although our DSKD method appears to be symplectic when tested with single particle integrations, it is not the best integrator for general N-body problems. We simply do not have a suitable Select operator that commutes with either the Drift or Kick operator, that is computationally cheap to evaluate compared to an acceleration, and that returns an accurate timestep in all cases. The local density criterion does well on the first count but only occasionally satisfies the second. This problem was foreshadowed by the inability to define the local density in the Kepler case and having to resort to a global criterion that used the enclosed density. When calculated in a cosmological simulation by averaging over the nearest $N$ particles, it erased substructure with $\lesssim N$ particles. We have not exhausted the search for an ideal timestep criterion, and there yet may be something that fits these stringent requirements. Nevertheless, the criterion we have explored in this paper may be very useful in simulations where structure is defined by large numbers of particles, such as in the modeling of galactic structure.

For more general cosmological N-body simulations we advocate either the SDKD or SKDK method using $\eta \sqrt{\epsilon/a}$ and/or (if clusters have significant substructure) $\eta(\epsilon/v)$ to choose the timestep. Using tests similar to those described for our symplectic integrator we determined that an appropriate choice for $\eta$ is 0.3 (for Plummer softening 0.4) to insure stability of the integration. Although this algorithm is not symplectic, it appears to give accurate results for quantities of interest in a cosmological N-body simulation.

This work was supported in part by NASA HPCC/ESS grant NAG 5-2213 and by NASA Astrophysics Theory grant NAGW-2523. We also wish to acknowledge useful discussions with Scott Tremaine.
A. Symplectic Integrators in Comoving Coordinates

The equations of motion in a comoving coordinate frame are traditionally presented as
\[
\dot{v'} + 2H(t)v' = -\frac{\nabla' \phi'}{a^3} \\
\dot{r'} = v' \\
\nabla'^2 \phi' = 4\pi G(\rho' - \rho_b'),
\]
where the primed quantities refer to the comoving particle motion in the comoving frame is symplectic. The Lagrangian for the particle motion in comoving coordinates is
\[
L = \frac{1}{2}(av' + \dot{a}r')^2 - \phi.
\]

With the (time dependent) generating function $\psi = (1/2)a\dot{a}r'^2$, this transforms to
\[
\mathcal{L} = \frac{1}{2}a^2v'^2 - \phi'/a,
\]
where
\[
\phi' = a\phi + \frac{1}{2}\ddot{a}a^2r'^2.
\]

(See Peebles, 1980.) However, these equations do not appear to be symplectic, i.e., there is no known generating function that produces this transformation.

However, by making a suitable canonical transformation, one can easily derive an integrator that is symplectic. The Lagrangian for the particle motion in the comoving frame is
\[
H = \frac{p'^2}{2a^2} + \frac{\phi'}{a}.
\]

Although this Hamiltonian is time-dependent\(^4\) it is separable, so the “drift” and “kick” operators are easily derived as:
\[
D(\tau) \equiv \int_t^{t+\tau} dt = 2 \int \frac{a^{-1/2}(t) - a^{-1/2}(t + \tau)}{H_0} dt,
\]
\[
K(\tau) = \int_t^{t+\tau} dt = 2 \int \frac{a^{1/2}(t + \tau) - a^{1/2}(t)}{H_0} dt.
\]

For non-flat matter dominated universes, it is convenient to use the parametric solutions (Peebles, 1980),
\[
a = A(1 - \cos \eta), \quad t = B(\eta - \sin \eta); \quad \Omega > 1
\]
\[
a = A(\cosh \eta - 1), \quad t = B(\sinh \eta - \eta); \quad \Omega < 1,
\]
where the constants $A$ and $B$ are
\[
A = 4/3\pi G\rho_b a^3|R|^2, \quad B = A|R|,
\]
$R$ is the curvature radius and the integrals are
\[
\int_t^{t+\tau} dt = \frac{B}{A}[\eta(t+\tau) - \eta(t)],
\]
and
\[
\int_t^{t+\tau} dt = \frac{B}{A^2} \left[ \cot \frac{\eta(t)}{2} - \cot \frac{\eta(t+\tau)}{2} \right]; \quad \Omega > 1
\]
\[
= \frac{B}{A^2} \left[ \coth \frac{\eta(t)}{2} - \coth \frac{\eta(t+\tau)}{2} \right]; \quad \Omega < 1.
\]

The above operators can be used to build leapfrog or higher order symplectic integrators for particle motion in comoving coordinates.

\(^4\)An energy can still be defined, as in the Layzer-Irvine energy equation. See Peebles, 1980.
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