# COSC3500 2D Orbital Simulation Report

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## Description

The task was to create a stock-standard, 2-dimensional gravitational n-body simulator. All bodies were to be assumed to be point masses. The simulation was to be accurate, maintaining a constant total energy, and exhibiting phenomena such as apsidal precession. The simulation was to accept arguments specifying the granularity of the simulation (number of time steps, number of instances of data export), and a file specifying masses and their initial positions and velocities. The simulation was to produce its output as fast as possible, with minimal slowdown with an increasing n number of bodies.

The simulator did not need to handle collisions between bodies. The MS1 simulator was to be free of OpenMP multiprocessing, which would be implemented in preparation for the MS2 submission.

## Implementation

At a high-level, the initial naive simulator:

- 1. Parsed input parameters and files
- 2. Constructed Body class instances, representing each point mass
- 3. Packed the Bodys into a std::vector<Body>, to maximize cache locality
- 4. Calculated forces between all pairwise combinations of n-bodies  $(O(n^2))$
- 5. Performed Euler's method to derive new velocities and positions
- 6. Output all necessary data
- 7. GOTO 4

By using a Quadtree, ('a tree datastructure in which each internal node has exactly four children') and the Barnes-Hut algorithm[1], the total cost of force calculation could be reduced to  $O(n \log n)$ , by grouping close-together bodies and approximating forces between the singular grouped pseudo-body, and distant bodies. New Quadtrees were constructed on each separate simulation step.

Dehen and Read note that the Euler method 'performs very poorly in practice', further noting that 'errors are proportional to  $\Delta t^2$ '. They contrast it with the second-order leapfrog symplectic integrator, which is 'heavily used in collisionless N-body applications'.[2]

Leapfrog can be expressed many in forms[3] including a synchronised form:

$$x_{i} = x_{i-1} + v_{i-1/2} \Delta t$$

$$a_{i} = F(x_{i})$$

$$v_{i+1/2} = v_{i-1/2} + a_{i} \Delta t$$

which only requires a single acceleration calculation per every two half timesteps (the timestep  $\Delta t$  must be constant to maintain stability), and a 'kick-drift-kick' form

$$v_{i+1/2} = v_i + a_i \frac{\Delta t}{2}$$

$$x_{i+1} = x_i + v_{i+1/2} \Delta t$$

$$v_{i+1} = v_{i+1/2} + a_{i+1} \frac{\Delta t}{2}$$

that is stable with variable timstepping, but incurs an additional acceleration calculation per every two half timesteps.

The synchronised form was implemented, but attempts to implement the kick-drift-kick form, and variable timestepping, were left unfinished.

Thus, the final implementation:

- 1. Parsed input parameters and files
- 2. Constructed Body class instances, representing each point mass
- 3. Packed the Bodys into a std::vector<Body>, to maximize cache locality
- 4. Inserted all Bodys into a fresh QuadTree on full timesteps, traversing the QuadTree with every Body to calculate forces  $(O(n \log n))$
- 5. Performed the appropriate Leapfrog step to derive new velocities or positions
- 6. Output all necessary data
- 7. GOTO 4

#### Correctness

I personally believe that the simulation is mostly right. By visualising the results with matplotlib, we see something that resembles an n-body simulator<sup>1</sup>. Distressingly, the total energy of the system is not constant throughout the system. When bodies are in close proxmity, anomalous energies that are inconsistent with the energy curve are observed. Furthermore, simulations with higher numbers of bodies produce random low energy outliers, but have a smoother total energy curve. There seems to be no significant difference between Euler method and Leapfrog, with respect to energy anomalies.

<sup>&</sup>lt;sup>1</sup>and that's good enough for me

## Performance & Scaling

Henceforth, the use of the 'recognizable' refers to eyeballing the output data, and making no significant effort to investigate the data more rigorously.

The addition of the -march=native compiler flag, which enables the use of all CPU specific instructions, provided no recognizable improvement in running time, but was left enabled in the instance that it improved performance on goliath.

The use of both th GCC and Clang Profile-Guided Optimisation features provided no recognizable improvement in running time.

Distressingly, -00, -01, -02, -03 showed no recognizable improvement in running time. -0fast led to an -4%-ish performance regression.

By profiling with callgrind, we saw that execution was dominated by only one user-defined method.

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Incl.		Self		Called	Function	Location
	100.00		0.00	(0)	■ 0x00000000001030	ld-2.17.so
	100.00		0.00	2	_dl_runtime_resolve_xsave	ld-2.17.so
	100.00		0.00	1	0x0000000000401df6	nbody
	100.00		0.00	(0)	(below main)	libc-2.17.so
	100.00	1	9.33	1	main	nbody: main.cpp, basic_string.h, string_conversio
	89.22		26.33	672 000 056	■ Body::exert_force_unidirec	nbody: Body.cpp
	62.90		2.93	672 033 712	distance(double, double,	nbody: utils.cpp
	59.97	1	11.70	672 033 711	■ hypot	libm-2.17.so
	48.27		48.27	672 033 712	hypot_finite	libm-2.17.so
1	0.63		0.63	48 000 000	Body::frog(double)	nbody: Body.cpp
1	0.52		0.52	48 000 008	Body::leap(double)	nbody: Body.cpp
	0.21		0.21	48 000 000	Body::reset_force()	nbody: Body.cpp
	0.07		0.00	601	dump_timestep(double, st	nbody: main.cpp, stl_vector.h, stl_iterator.h

Two performance fixes were divised.

Here, we are recalculating

gcc -g -lstdc++ -Wall -pedantic -Wextra -std=c++11 -lm -O3 -march=native Body.o QuadTree.o main.o utils.o -o nbody Barnes-Hut enabled: false Leapfrog enabled: true Total CPU time was  $13.938825\ 12000001$  simulation steps computed

### References

- [1] J. E. Barnes and P. Hut, "A hierarchical O(n-log-n) force calculation algorithm," *Nature*, vol. 324, p. 446, 1986.
- [2] W. Dehnen and J. I. Read, "N-body simulations of gravitational dynamics," *European Physical Journal Plus*, vol. 126, p. 55, May 2011.
- [3] R. D. Skeel, "Variable step size destabilizes the störmer/leapfrog/verlet method," BIT Numerical Mathematics, vol. 33, pp. 172–175, Mar 1993.

Figure 1: Observed energy anomaly while bodies in close proxmity - Leapfrog

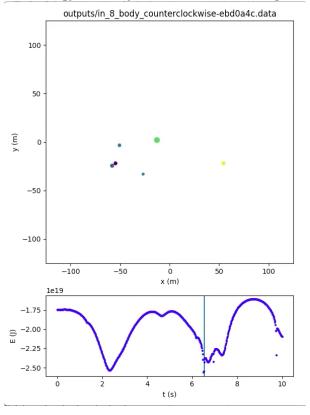


Figure 2: Observed energy anomaly while bodies in close proxmity - Euler method

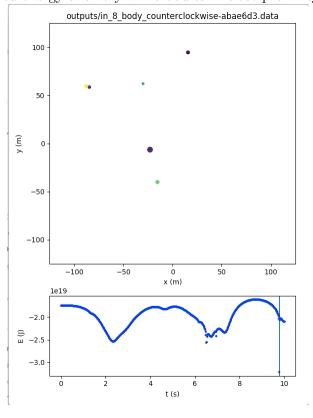


Figure 3: Observed energy anomaly while bodies in close proxmity - Euler method

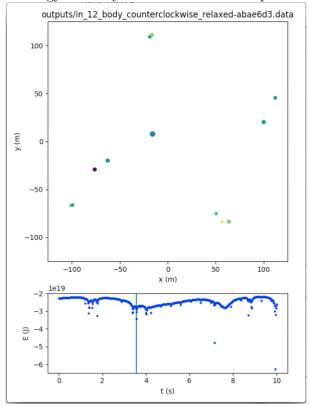


Figure 4: Observed energy anomaly - random outlier - Leapfrog

