Computational Physics Lab

Homework 3

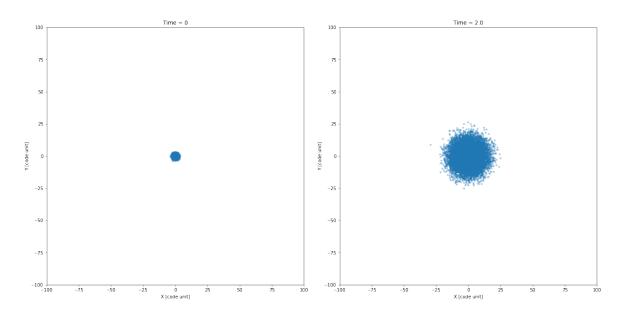
108000204 Yuan-Yen Peng Dept. of Physics, NTHU Hsinchu, Taiwan

December 4, 2022

1 Programming Assignments

1.1 (1) Normal Cloud

In this section, in the beginning, we create a cloud of particles ($N=10^4$) distributed by a normal distribution with zero mean and one variance (or in other words, s.d.=1) in 3D Cartesian coordinates. Additionally, we set the initial particle velocities, and accelerations are distributed by the same normal distribution and the system's total mass is 20. We, afterward, use our N body simulation code from t = 0 up to t = 10 with a constant time step $\Delta t=0.01$ and a soften length $r_{soft}=0.01$. We, meanwhile, implement numba @njit with 8 cores to accelerate the loop speed (@jit (nopython=True) have spent more time so I discard to use it). Besides, I, firstly, set the particle's number = 10^5 tried to use either @jit or @njit to accelerate the code; however, with RK4 it takes no lowering than 800 minutes! Therefore, I use 10^4 , instead. Figure 1 is the group of snap-shot of the normal cloud distribution with the RK4 method and with the extended boundary from -50 to +50. This is a time-consuming task, for orders of 4, RK4, it took 42015 [s]; for orders of two, RK2 and Leapfrog, they took 29506 and 18009 [s], respectively. The last order of one, Euler, took 17794 [s].



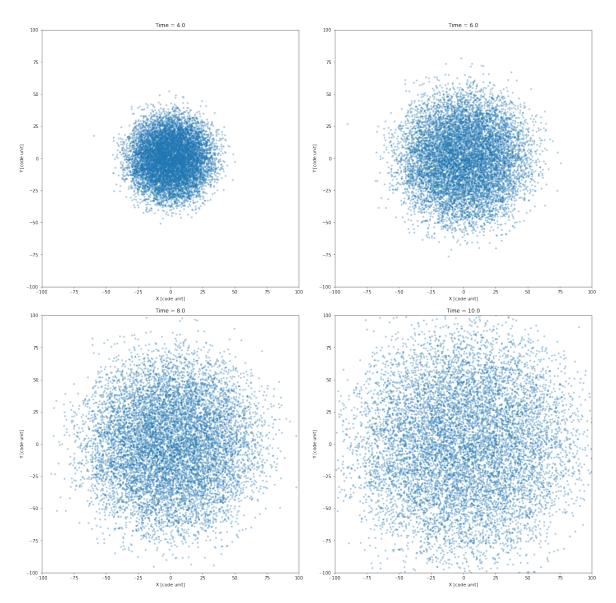


Figure 1: This the normal cloud with RK4, from t = 0 to t = 10 and dt = 0.01. One run took: 18009 [s]

1.2 (2) Leapfrog method

The leapfrog method is a second-order method for solving an initial value problem. The main idea in this is divided into three parts, kick-drift-kick. For each time step dt, each particle receives a half-step kick,

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

and use the above half-step velocity, and followed by a full-step drift,

$$x_{i+1} = x_i + v_{i+1/2}dt$$

and lastly, use the above drift, and followed by another half-step kick,

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

In the next subsection, we will discuss and compare the Leapfrog method with Euler, RK2, and RK4 as well.

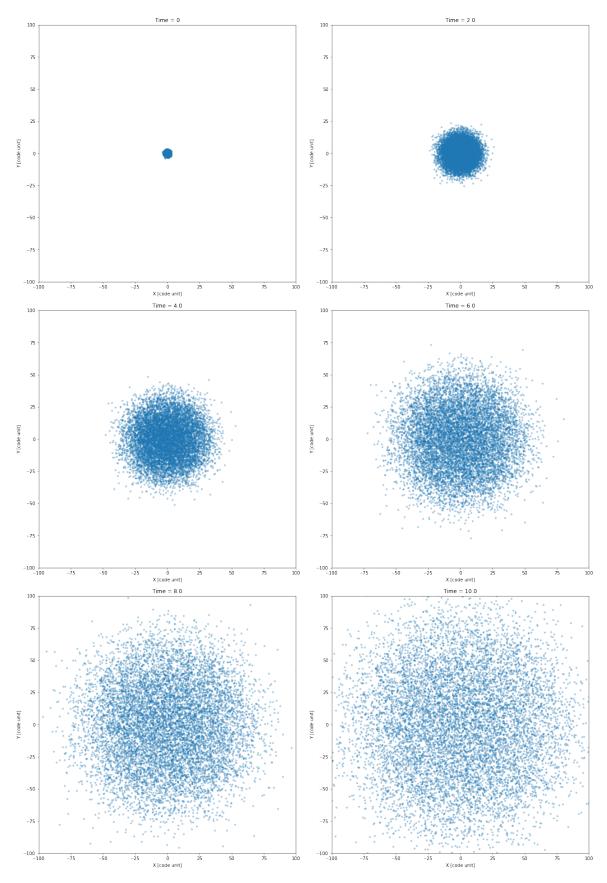


Figure 2: This the normal cloud with Leapfrog, from t = 0 to t = 10 and dt = 0.01. One run took: 18009 [s]

1.3 (3) Energy comparison

In this section, we will discuss kinetic energy, potential energy, and total energy. The definitions are below:

$$U = -\frac{Gm_im_j}{r + r_{soft}}$$
$$K = \frac{1}{2}m_iv_i^2$$
$$E = U + K$$

where the subscriptions indicate which particles; G is the gravitational constant, here we set it to 1; m is the mass of each particle; r and r_{soft} is the distance between two particles and the tolerance of distance, which can avoid the nominator equal to zero. After defining all the parameters, we use four different methods, Euler, RK2, RK4, and Leapfrog, and implement the initial conditions as in section 1 to rerun the codes.

See in Figure 3~6, we can find out that the potential energy is much smaller than the kinetic energy owing to the small masses of particles. On the other hand, the total energy is dominated by the kinetic part, which makes sense to us. In Figure 3~6, we can also see the total energy is not the "constant" at the beginning because we initially set the normal random accelerations to each particle which can be regarded that there are other forces applied to the particles. Nevertheless, in the wake of a little period, the total energy will gradually be in a stable stage because of no external force. These phenomena not merely can see in the Euler but in others as well. Compare this with the stable speed, which means how long to be stable. We use RK4 as the theoretical benchmark. It is obvious that Euler (first order) is the slowest one; in the other words, it is the most distorted one; for the second order, RK2 is shapeless at the beginning, and Leapfrog is the most similar one with the benchmark. Therefore, we can conclude that if individuals want to calculate a large number of particles problem, they can choose the Leapfrog method because it has not merely high speed but also high accuracy. (detail of accuracy can see in the next section, the accuracy compared with the benchmark is within one order.)

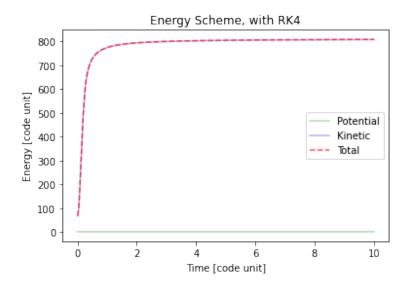


Figure 3: RK4, y-axis is Energy; x-axis is time.

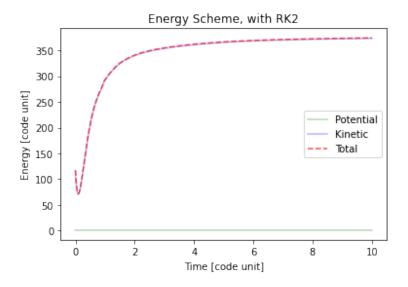


Figure 4: RK2, y-axis is Energy; x-axis is time.

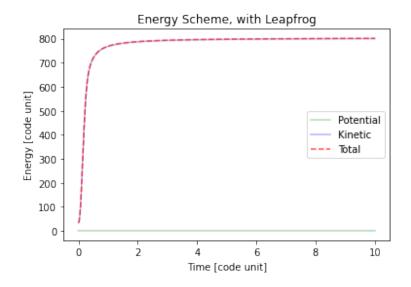


Figure 5: Leapfrog, y-axis is Energy; x-axis is time.

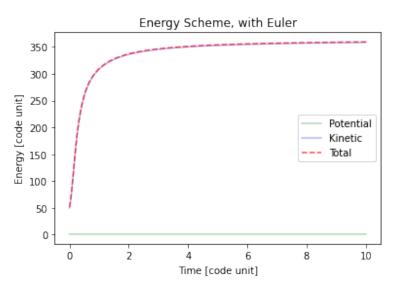


Figure 6: Euler, y-axis is Energy; x-axis is time.

So as to analyze the accuracy of the Leapfrog method, we set the RK4 (fourth order) as a benchmark. We, likewise, applied a logarithm of 10 to investigate the order of accuracy thereafter. In Figure 7, we can see the accuracy comparisons between Leapfrog methods and the benchmark.

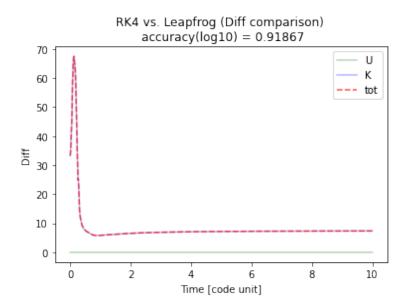


Figure 7: In the comparison of energy schemes between the RK4 and Leapfrog methods, the average accuracy is approximately 0.9 orders.

2 Codes

All the codes are transferred from jupyterlab or python codes; hence, if you want to re-run them, please see the source code in the attached files or my GitHub repository:

<https://github.com/gary20000915/Comphyslab-HW3.git>

2.1 particles.py

```
import numpy as np
   class Particles:
      0.00
      The Particles class handle all particle properties
      for the N-body simulation.
10
      def __init__(self, N:int = 100):
          Prepare memories for N particles
14
          :param N: number of particles.
15
          By default: particle properties include:
17
                 nparticles: int. number of particles
18
                 _masses: (N,1) mass of each particle
19
```

```
_positions: (N,3) x,y,z positions of each particle
20
                 _velocities: (N,3) vx, vy, vz velocities of each particle
21
                 _accelerations: (N,3) ax, ay, az accelerations of each partciel
                 _tags: (N) tag of each particle
                 _time: float. the simulation time
          0.00
26
          self.nparticles = N
          self._time = 0 # initial time = 0
          self._masses = np.ones((N, 1))
          self._positions = np.zeros((N, 3))
          self._velocities = np.zeros((N, 3))
31
          self._accelerations = np.zeros((N, 3))
          self._tags = np.linspace(1, N, N)
          self._U = np.ones((N, 1))
          self._K = np.ones((N, 1))
          return
38
39
      def get_time(self):
40
          return self._time
41
      def get_masses(self):
43
          return self._masses
45
      def get_positions(self):
46
          return self._positions
      def get_velocities(self):
          return self._velocities
50
51
      def get_accelerations(self):
          return self._accelerations
      def get_tags(self):
55
          return self._tags
56
57
      def get_time(self):
58
          return self._time
60
      def get_U(self):
61
          return self._U
62
63
      def get_K(self):
          return self._K
67
      def set_time(self, time):
68
          self. time = time
69
          return
70
71
      def set_masses(self, mass):
72
```

```
self._masses = mass
           return
74
75
       def set_positions(self, pos):
76
           self._positions = pos
           return
79
       def set_velocities(self, vel):
80
           self._velocities = vel
81
           return
82
       def set_accelerations(self, acc):
84
           self._accelerations = acc
           return
86
87
       def set_tags(self, IDs):
           self._tags = IDs
           return
91
       def set_U(self, U):
92
           self._U = U
93
           return
       def set_K(self, K):
96
           self._K = K
97
           return
98
99
       def output(self, fn, time):
100
           Write simulation data into a file named "fn"
102
103
           mass = self._masses
104
           pos = self._positions
105
           vel = self._velocities
           acc = self._accelerations
           tag = self._tags
108
           header = """
109
                  Data from a 3D direct N-body simulation.
111
112
                  rows are i-particle;
113
                  coumns are :mass, tag, x ,y, z, vx, vy, vz, ax, ay, az
114
                  NTHU, Computational Physics Lab
116
117
118
119
           header += "Time = {}".format(time)
120
           np.savetxt(fn,(tag[:],mass[:,0],pos[:,0],pos[:,1],pos[:,2],
                             vel[:,0],vel[:,1],vel[:,2],
                             acc[:,0],acc[:,1],acc[:,2]),header=header)
123
124
           return
```

2.2 simulation.py

```
import numpy as np
  from pathlib import Path
   import time
   from numba import jit, njit, prange, set_num_threads
  from nbody.particles import Particles
   import matplotlib.pyplot as plt
   0.00
8
  This program solve 3D direct N-particles simulations
10
   under gravitational forces.
11
  This file contains two classes:
13
14
   1) Particles: describes the particle properties
15
   2) NbodySimulation: describes the simulation
16
  Usage:
18
19
      Step 1: import necessary classes
20
21
      from nbody import Particles, NbodySimulation
      Step 2: Write your own initialization function
24
25
26
          def initialize(particles:Particles):
             particles.set_masses(mass)
30
             particles.set_positions(pos)
             particles.set_velocities(vel)
             particles.set_accelerations(acc)
33
             return particles
35
      Step 3: Initialize your particles.
37
38
          particles = Particles(N=100)
          initialize(particles)
40
42
      Step 4: Initial, setup and start the simulation
43
          simulation = Simulation(particles)
          simulation.setip(...)
          simulation.evolve(dt=0.001, tmax=10)
47
48
  Author: Yuan-Yen Peng (editted from Prof. Kuo-Chuan Pan, NTHU 2022.10.30)
  Dept. of Physics, NTHU
```

```
Date: Npv. 28, 2022
   For the course, computational physics lab
53
54
   0.00
55
   def ACC_norm(N, posx, posy, posz, G, mass, rsoft, U, K, vel_square):
58
       Acceleration with normal for loops.
59
60
       :param N: number of particles
61
       :param posx: position x
       :param posy: position y
63
       :param posz: position z
       :param G: gravitational constant
65
       :param mass: mass
66
       1 1 1
       acc = np.zeros((N, 3))
       for i in range(N):
70
          for j in range(N):
              if j > i:
                 x = posx[i] - posx[j]
73
                  y = posy[i] - posy[j]
                  z = posz[i] - posz[j]
75
                  r = np.sqrt(x**2 + y**2 + z**2) + rsoft
                  theta = np.arccos(z / r)
                  phi = np.arctan2(y, x)
78
                  F = -G * mass[i, 0] * mass[j, 0] / np.square(r)
                  Fx = F * np.cos(theta) * np.cos(phi)
                  Fy = F * np.cos(theta) * np.sin(phi)
                 Fz = F * np.sin(theta)
82
83
                  acc[i, 0] += Fx / mass[i, 0]
                  acc[j, 0] = Fx / mass[j, 0]
                  acc[i, 1] += Fy / mass[i, 0]
87
                  acc[j, 1] -= Fy / mass[j, 0]
88
89
                  acc[i, 2] += Fz / mass[i, 0]
90
                  acc[j, 2] = Fz / mass[j, 0]
91
92
                  U[i] = -G * mass[i, 0] * mass[j, 0] / r
93
                  K[i] = 0.5 * (mass[i, 0] * np.sum(vel_square[i, :])
94
                               + mass[j, 0] * np.sum(vel_square[j, :]))
95
       return acc, U, K
96
97
   @jit(nopython=True)
   def ACC_jit(N, posx, posy, posz, G, mass, rsoft, U, K, vel_square):
99
100
       Acceleration with numba jit for loops
101
102
       :param N: number of particles
103
       :param posx: position x
```

```
105
       :param posy: position y
       :param posz: position z
106
       :param G: gravitational constant
107
       :param mass: mass
108
       acc = np.zeros((N, 3))
       for i in range(N):
111
          for j in range(N):
              if j > i:
                  x = posx[i] - posx[j]
114
                  y = posy[i] - posy[j]
115
                  z = posz[i] - posz[j]
                  r = np.sqrt(x**2 + y**2 + z**2) + rsoft
                  theta = np.arccos(z / r)
118
                  phi = np.arctan2(y, x)
119
                  F = -G * mass[i, 0] * mass[j, 0] / np.square(r)
120
                  Fx = F * np.cos(theta) * np.cos(phi)
                  Fy = F * np.cos(theta) * np.sin(phi)
                  Fz = F * np.sin(theta)
124
                  acc[i, 0] += Fx / mass[i, 0]
                  acc[j, 0] = Fx / mass[j, 0]
126
                  acc[i, 1] += Fy / mass[i, 0]
                  acc[j, 1] = Fy / mass[j, 0]
130
                  acc[i, 2] += Fz / mass[i, 0]
                  acc[j, 2] = Fz / mass[j, 0]
133
                  U[i] = -G * mass[i, 0] * mass[j, 0] / r
                  K[i] = 0.5 * (mass[i, 0] * np.sum(vel_square[i, :])
                               + mass[j, 0] * np.sum(vel_square[j, :]))
136
       return acc, U, K
138
   set_num_threads(8)
   @njit(parallel=True)
140
   def ACC_njit(N, posx, posy, posz, G, mass, rsoft, U, K, vel_square):
141
       1.1.1
142
       Acceleration with numba njit for loops (parallel)
143
       :param N: number of particles
145
       :param posx: position x
146
       :param posy: position y
147
       :param posz: position z
148
       :param G: gravitational constant
149
       :param mass: mass
150
       acc = np.zeros((N, 3))
152
       for i in prange(N):
          for j in prange(N):
              if j > i:
155
                  x = posx[i] - posx[j]
156
                  y = posy[i] - posy[j]
157
```

```
z = posz[i] - posz[j]
158
                  r = np.sqrt(x**2 + y**2 + z**2) + rsoft
159
                  theta = np.arccos(z / r)
160
                  phi = np.arctan2(y, x)
161
                  F = -G * mass[i, 0] * mass[j, 0] / np.square(r)
162
                  Fx = F * np.cos(theta) * np.cos(phi)
                  Fy = F * np.cos(theta) * np.sin(phi)
164
                  Fz = F * np.sin(theta)
165
166
                   acc[i, 0] += Fx / mass[i, 0]
167
                   acc[j, 0] = Fx / mass[j, 0]
168
169
                   acc[i, 1] += Fy / mass[i, 0]
170
                  acc[j, 1] = Fy / mass[j, 0]
                   acc[i, 2] += Fz / mass[i, 0]
173
                   acc[j, 2] = Fz / mass[j, 0]
174
                  U[i] = -G * mass[i, 0] * mass[j, 0] / r
176
                  K[i] = 0.5 * (mass[i, 0] * np.sum(vel_square[i, :])
                                + mass[j, 0] * np.sum(vel_square[j, :]))
178
       return acc, U, K
179
181
   class NbodySimulation:
       11 11 11
182
183
       The N-body Simulation class.
184
       0.00
187
       def __init__(self,particles:Particles):
188
189
           Initialize the N-body simulation with given Particles.
190
           :param particles: A Particles class.
193
           0.00
194
195
           # store the particle information
196
           self.nparticles = particles.nparticles
           self.particles = particles
198
199
           # Store physical information
200
           self.time = 0.0 # simulation time
201
202
           # Set the default numerical schemes and parameters
           self.setup()
205
           return
206
207
       def setup(self, G=1,
208
                      rsoft=0.01,
209
                      method="RK2",
210
```

```
io_freq=10,
                      io_title="particles",
212
                      io_screen=True,
                      visualized=False):
214
           0.00
           Customize the simulation environments.
           :param G: the graivtational constant
218
           :param rsoft: float, a soften length
           :param meothd: string, the numerical scheme
                         support "Euler", "RK2", and "RK4"
221
           :param io_freq: int, the frequency to outupt data.
223
                          io_freq <=0 for no output.</pre>
224
           :param io_title: the output header
225
           :param io_screen: print message on screen or not.
226
           :param visualized: on the fly visualization or not.
           0.00
229
           # TODO:
230
           self.G = G
           self.rsoft = rsoft
           self.method = method
           self.io_freq = io_freq
234
           self.io_title = io_title
235
           self.io_screen = io_screen
236
           self.visualized = visualized
           return
239
       def evolve(self, dt:float=0.01, tmax:float=1):
240
           0.00
241
242
           Start to evolve the system
243
           :param dt: time step
           :param tmax: the finial time
246
247
           0.00
248
           # TODO:
249
           method = self.method
           if method=="Euler":
251
               _update_particles = self._update_particles_euler
252
           elif method=="RK2":
253
               _update_particles = self._update_particles_rk2
254
           elif method=="RK4":
255
               _update_particles = self._update_particles_rk4
           elif method=="Leapfrog":
              _update_particles = self._update_particles_lf
258
           else:
259
              print("No such update meothd", method)
260
              quit()
261
262
           # prepare an output folder for lateron output
263
```

```
io_folder = "data_"+self.io_title
264
          Path(io_folder).mkdir(parents=True, exist_ok=True)
265
          io_folder_fig = "fig_" + self.io_title
266
          Path(io_folder_fig).mkdir(parents=True, exist_ok=True)
267
          270
          # The main loop of the simulation
271
          273
          # TODO:
275
          particles = self.particles # call the class: Particles
276
          t = self.particles.get_time()
278
          t1 = time.time()
279
          step = int(tmax / dt) + 1
          UU = np.ones((step, 1))
          KK = np.ones((step, 1))
282
          EE = np.ones((step, 1))
283
          for i in range(step):
284
             # update particles
285
             particles = _update_particles(dt, particles)[0]
             UU[i] = _update_particles(dt, particles)[1]
             KK[i] = _update_particles(dt, particles)[2]
288
             EE[i] = UU[i] + KK[i]
289
             # update io
290
             if (n % self.io_freq == 0):
291
                 if self.io_screen:
                    # print('n = ', n, 'time = ', t, 'dt = ', dt)
293
                    # output
294
                    fn = io_folder+"/data_"+self.io_title+"_"+str(n).zfill(5)+".txt"
295
                    print(fn)
296
                    self.particles.output(fn, t)
                    # savefig
299
                    scale = 100
300
                    fig, ax = plt.subplots()
301
                    fig.set_size_inches(10.5, 10.5, forward=True)
302
                    fig.set_dpi(72)
303
                    ax.set_xlim(-1*scale,1*scale)
304
                    ax.set_ylim(-1*scale,1*scale)
305
                    ax.set_aspect('equal')
306
                    ax.set_xlabel('X [code unit]')
307
                    ax.set_ylabel('Y [code unit]')
308
                    pos = particles.get_positions()
                    plt.title(f'Time = {np.round(t, 0)}')
                    FIG = f'{io_folder_fig}/fig_{self.io_title}_{str(int(0.01 *
311
                        n)).zfill(1)}.png'
                    ax.scatter(pos[:, 0], pos[:, 1], s = 10, alpha = .3)
312
                    plt.savefig(FIG)
313
                    plt.show()
315
```

```
# update time
              if t + dt > tmax:
317
                  dt = tmax - t
318
              t += dt
319
              n += 1
320
           T = np.linspace(0, tmax, step)
322
           plt.plot(T, UU, 'g', alpha = .3, label = 'Potential')
           plt.plot(T, KK, 'b', alpha = .3, label = 'Kinetic')
324
           plt.plot(T, EE, '--r', alpha = .7, label = 'Total')
325
           plt.xlabel('Time [code unit]')
326
           plt.ylabel('Energy [code unit]')
           plt.title(f'Energy Scheme, with {method}')
           plt.legend()
329
           plt.show()
330
           t2 = time.time()
           print("Time diff: ", t2 - t1)
           print("Done!")
334
           return [UU, KK]
335
336
       def _calculate_acceleration(self, mass, pos):
           0.00
           Calculate the acceleration.
339
           0.00
340
           # TODO:
341
           particles = self.particles
342
           G = self.G
           rsoft = self.rsoft
           posx = pos[:, 0]
345
           posy = pos[:, 1]
346
           posz = pos[:, 2]
347
           N = self.nparticles
348
           U = particles.get_U()
           K = particles.get_K()
           vel_square = np.square(particles.get_velocities())
351
           # return ACC_jit(N, posx, posy, posz, G, mass,rsoft, U, K, vel_square)
352
           return ACC_njit(N, posx, posy, posz, G, mass, rsoft, U, K, vel_square)
354
       def _update_particles_euler(self, dt, particles:Particles):
355
           # TODO:
356
           mass = particles.get_masses()
357
           pos = particles.get_positions()
358
           vel = particles.get_velocities()
359
           acc = self._calculate_acceleration(mass, pos)[0]
360
           y0 = np.array([pos, vel])
361
           yder = np.array([vel, acc])
363
           y0 = np.add(y0, yder * dt)
364
365
           U = np.sum(self._calculate_acceleration(mass, y0[0])[1])
           K = np.sum(self._calculate_acceleration(mass, y0[0])[2])
367
368
```

```
particles.set_positions(y0[0])
369
          particles.set_velocities(y0[1])
          particles.set_accelerations(acc)
371
          return particles, U, K
373
       def _update_particles_rk2(self, dt, particles:Particles):
375
          # TODO:
376
          mass = particles.get_masses()
          pos = particles.get_positions()
378
          vel = particles.get_velocities()
379
          acc = self._calculate_acceleration(mass, pos)[0]
          y0 = np.array([pos, vel])
382
          yder = np.array([vel, acc])
383
          k1 = yder
384
          y_{temp} = y0 + dt * k1
          acc = self._calculate_acceleration(mass, y_temp[0])[0]
          k2 = np.array([y_temp[1], acc])
387
          y0 = np.add(y0, (dt / 2) * (k1 + k2))
388
389
          acel = self._calculate_acceleration(mass, y0[0])[0]
390
          U = np.sum(self._calculate_acceleration(mass, y0[0])[1])
          K = np.sum(self._calculate_acceleration(mass, y0[0])[2])
392
393
          particles.set_positions(y0[0])
394
          particles.set_velocities(y0[1])
395
          particles.set_accelerations(acel)
396
          return particles, U, K
398
       def _update_particles_rk4(self, dt, particles:Particles):
399
400
          mass = particles.get_masses()
401
          pos = particles.get_positions()
402
          vel = particles.get_velocities()
          acc = self._calculate_acceleration(mass, pos)[0]
404
405
          y0 = np.array([pos, vel])
406
          yder = np.array([vel, acc])
407
          k1 = yder
408
          y_{temp} = y0 + 0.5 * dt * k1
409
          acc = self._calculate_acceleration(mass, y_temp[0])[0]
410
          k2 = np.array([y_temp[1], acc])
411
          y_{temp} = y0 + 0.5 * dt * k2
412
          acc = self._calculate_acceleration(mass, y_temp[0])[0]
413
          k3 = np.array([y_temp[1], acc])
          y_temp = y0 + dt * k3
          acc = self._calculate_acceleration(mass, y_temp[0])[0]
416
          k4 = np.array([y_temp[1], acc])
417
418
          y0 = np.add(y0, (1/6) * dt * (k1 + 2*k2 + 2*k3 + k4))
419
420
          acel = self._calculate_acceleration(mass, y0[0])[0]
421
```

```
U = np.sum(self._calculate_acceleration(mass, y0[0])[1])
          K = np.sum(self._calculate_acceleration(mass, y0[0])[2])
423
424
          particles.set positions(y0[0])
425
          particles.set_velocities(y0[1])
          particles.set_accelerations(acel)
          return particles, U, K
428
429
       def _update_particles_lf(self, dt, particles:Particles):
430
          # TODO:
431
          mass = particles.get_masses()
          pos = particles.get_positions()
433
          vel = particles.get_velocities()
435
          acc = particles.get_accelerations()
436
          vel_prime = vel + acc * 0.5 * dt
          pos = pos + vel_prime * dt
438
          acc = self._calculate_acceleration(mass, pos)[0]
          vel = vel_prime + acc * 0.5 * dt
440
441
          U = np.sum(self. calculate acceleration(mass, pos)[1])
442
          K = np.sum(self._calculate_acceleration(mass, pos)[2])
          particles.set_positions(pos)
445
          particles.set_velocities(vel)
          particles.set_accelerations(acc)
447
          return particles, U, K
448
450
   if __name__=='__main__':
451
452
       # test Particles() here
453
       particles = Particles(N=10)
       # test NbodySimulation(particles) here
       sim = NbodySimulation(particles=particles)
       sim.setup(G = 6.67e-8, io_freq=2, io_screen=True, io_title="test")
457
       sim.evolve(dt = 1, tmax = 10)
458
       print(sim.G)
459
       print("Done")
```

2.3 NormalCloud.ipynb

```
# # %% [markdown]
# ## Homework 3
# ### Programming Assignment 1
# ### 111 Computational Physics Lab
# >Author: Yuan-Yen Peng 108000204
# >Email: garyphys0915@gapp.nthu.edu.com
# >Date: Nov. 11, 2022
# >LICENCE: MIT
# %%
```

```
import numpy as np
  import glob
12
  from numba import jit, njit, prange, set_num_threads
13
   import matplotlib.pyplot as plt
   from mpl_toolkits.mplot3d import Axes3D
   import matplotlib.animation as animation
   from nbody.particles import Particles
  from nbody.simulation import NbodySimulation
18
19
  # %%
  problem_name = "NormalCloud"
21
  Num = int(1e4)
  tmax = 10
  dt = 0.01
  r_soft = 0.001
25
  # %%
27
  set_num_threads(8)
   @njit(parallel = True)
   def generator(N, positions, velocities, accelerations):
30
          mu, sigma = 0, 1 # mean = 0; variance = 1, i.e., standard deviation =
31
             sqrt(var) = 1
          for i in prange(N):
             positions[i] = np.random.normal(mu, sigma, 3)
33
             velocities[i] = np.random.normal(mu, sigma, 3)
             accelerations[i] = np.random.normal(mu, sigma, 3)
36
          return [positions, velocities, accelerations]
  # %%
30
   def initialRandomParticles(N):
40
41
          Initial particles
42
          0.00
          total_mass = 20
45
          particles = Particles(N = N)
46
47
          positions = particles.get_positions()
48
          velocities = particles.get_velocities()
          accelerations = particles.get_accelerations()
50
          masses = particles.get_masses() # ones array (size = N)
51
          mass = total_mass / particles.nparticles # single particel's mass
52
53
          particles.set_masses((masses * mass))
          particles.set_positions(generator(N, positions, velocities,
             accelerations)[0])
          particles.set_velocities(generator(N, positions, velocities,
56
             accelerations)[1])
          particles.set_accelerations(generator(N, positions, velocities,
             accelerations)[2])
          return particles
59
```

```
# %% [markdown]
61
   # solve with t = 0 \sim 10 with dt = 0.01 and r_soft = 0.01.
62
   # %%
   # Initial particles here.
   method = "RK4"
   particles = initialRandomParticles(N = Num)
   # Run the n-body simulations
   sim = NbodySimulation(particles)
   sim.setup(G=1,method=method,io_freq=200,io_title=problem_name,io_screen=True,visualized=False)
   Energy_RK4 = sim.evolve(dt=dt,tmax=tmax)
72
   # %%
   # Initial particles here.
74
   method = "RK2"
   particles = initialRandomParticles(N = Num)
   # Run the n-body simulations
   sim = NbodySimulation(particles)
   sim.setup(G=1,method=method,io_freq=200,io_title=problem_name,io_screen=True,visualized=False)
   Energy_RK2 = sim.evolve(dt=dt,tmax=tmax)
80
81
   # %%
   # Initial particles here.
83
   method = "Euler"
   particles = initialRandomParticles(N = Num)
   # Run the n-body simulations
86
   sim = NbodySimulation(particles)
   sim.setup(G=1,method=method,io_freq=200,io_title=problem_name,io_screen=True,visualized=False)
   Energy_Euler = sim.evolve(dt=dt,tmax=tmax)
90
   # %%
91
   # Initial particles here.
   method = "Leapfrog"
   particles = initialRandomParticles(N = Num)
   # Run the n-body simulations
95
   sim = NbodySimulation(particles)
   sim.setup(G=1,method=method,io_freq=200,io_title=problem_name,io_screen=True,visualized=False)
97
   Energy_LF = sim.evolve(dt=dt,tmax=tmax)
98
   # %%
100
   TT = len(Energy_RK4[0])
101
   Time = np.linspace(0, tmax, len(Energy_RK4[0]))
102
   Diff_tot = np.abs((Energy_RK4[0] + Energy_RK4[1]) - (Energy_LF[0] + Energy_LF[1]))
103
   Diff = np.abs(np.array(Energy_RK4) - np.array(Energy_LF))
104
   avg = np.average(Diff_tot)
105
   accuracy = np.log10(avg)
   print("Average = ", avg)
107
   print("Accuracy (log) = ", accuracy)
108
   plt.plot(Time, Diff[0], "g", alpha = .3, label = "U")
   plt.plot(Time, Diff[1], "b", alpha = .3, label = "K")
   plt.plot(Time, Diff_tot, "--r", alpha = .7, label = "tot")
```

```
plt.xlabel("Time [code unit]")
   plt.ylabel("Diff")
   plt.title(f"RK4 vs. Leapfrog (Diff comparison) \n accuracy(log10) =
       {np.round(accuracy, 5)}")
   plt.legend()
116
   plt.show()
118
   # %%
119
   Energy_RK4 = Energy_RK2
120
   TT = len(Energy_RK4[0])
   Time = np.linspace(0, tmax, len(Energy_RK4[0]))
   Diff_tot = np.abs((Energy_RK4[0] + Energy_RK4[1]) - (Energy_LF[0] + Energy_LF[1]))
   Diff = np.abs(np.array(Energy_RK4) - np.array(Energy_LF))
124
   avg = np.average(Diff_tot)
125
   accuracy = np.log10(avg)
126
   print("Average = ", avg)
   print("Accuracy (log) = ", accuracy)
   plt.plot(Time, Diff[0], "g", alpha = .3, label = "U")
130
   plt.plot(Time, Diff[1], "b", alpha = .3, label = "K")
   plt.plot(Time, Diff_tot, "--r", alpha = .7, label = "tot")
   plt.xlabel("Time [code unit]")
   plt.ylabel("Diff")
   plt.title(f"RK2 vs. Leapfrog (Diff comparison) \n accuracy(log10) =
       {np.round(accuracy, 5)}")
   plt.legend()
136
   plt.show()
138
   # %%
139
   Energy_RK4 = Energy_Euler
140
   TT = len(Energy_RK4[0])
141
   Time = np.linspace(0, tmax, len(Energy_RK4[0]))
142
   Diff_tot = np.abs((Energy_RK4[0] + Energy_RK4[1]) - (Energy_LF[0] + Energy_LF[1]))
143
   Diff = np.abs(np.array(Energy_RK4) - np.array(Energy_LF))
   avg = np.average(Diff_tot)
   accuracy = np.log10(avg)
   print("Average = ", avg)
147
   print("Accuracy (log) = ", accuracy)
148
149
   plt.plot(Time, Diff[0], "g", alpha = .3, label = "U")
   plt.plot(Time, Diff[1], "b", alpha = .3, label = "K")
   plt.plot(Time, Diff_tot, "--r", alpha = .7, label = "tot")
152
   plt.xlabel("Time [code unit]")
   plt.ylabel("Diff")
154
   plt.title(f"Euler vs. Leapfrog (Diff comparison) \n accuracy(log10) =
       {np.round(accuracy, 5)}")
   plt.legend()
   plt.show()
```