```
# -*- coding: utf-8 -*-
2
3
    Created on Sun Jun 3 11:19:56 2018
5
    @author: Cemenenkoff
6
7
    import matplotlib
8
    import matplotlib.pyplot as plt
9
    import numpy as np
10 from numpy import linalg as LA
11
   plt.style.use('classic') #Use a serif font.
12
   from IPython.display import set matplotlib formats
13 set matplotlib formats('pdf', 'png')
14 plt.rcParams['savefig.dpi'] = 200
15 plt.rcParams['figure.autolayout'] = False
16 plt.rcParams['figure.figsize'] = 10, 6
   plt.rcParams['axes.labelsize'] = 16
17
   plt.rcParams['axes.titlesize'] = 20
18
19
   plt.rcParams['font.size'] = 10
20
  plt.rcParams['lines.linewidth'] = 2.0
21 plt.rcParams['lines.markersize'] = 6
22 plt.rcParams['legend.fontsize'] = 14
23 plt.rcParams['axes.facecolor'] = 'white'
24 plt.rcParams['savefig.facecolor']='white'
25
   matplotlib.rcParams['xtick.direction'] = 'out'
   matplotlib.rcParams['ytick.direction'] = 'out'
26
27
   #This import is necessary for the isometric plot.
28
   from mpl toolkits.mplot3d import Axes3D
29
   31
32
   #Choose an integer number of bodies to explore between 2 and 10.
33
34
   #Choose a stepping method. Choices are:
35
   # 'euler' (Euler)
36
      'ec'
            (Euler-Cromer)
37
     'rk2' (2nd Order Runge-Kutta)
     'VV'
38
            (velocity Verlet)
     'pv' (position Verlet)
39
       'vefrl' (velocity extended Forest-Ruth-like)
40
    #
41
       'pefrl' (position extended Forest-Ruth-like)
42
   method = 'euler'
43
44
   #Set time parameters for the simulation.
45 t0 = 0.0 #start time in years
46
   tf = 20.0 #final time in years
47
   dt = 2.0 #time step in hours
48
   49
   50
   51
52
   #Define the masses for a number of bodies (in kg).
m0 = 1.989e30 \# mass of the sun
m1 = 3.285e23 \# mass of Mercury
55
  m2 = 4.867e24 \# mass of Venus
m3 = 5.972e24 \# mass of Earth
m4 = 6.417e23 \# mass of Mars
m5 = 1.898e27 \# mass of Jupiter
m6 = 5.683e26 \# mass of Saturn
m7 = 8.681e25 \# mass of Uranus
61
   m8 = 1.024e26 \# mass of Neptune
   m9 = 1.309e22 \# mass of Pluto
62
63
   #Combine the masses into a global masses list.
64
   m = [m0, m1, m2, m3, m4, m5, m6, m7, m8, m9]
6.5
66
    #Create a list of colors that is at least as long as the mass list so each
    #gravitationally interacting body has its own color.
```

```
c0 = '#ff0000' #red
     c1 = '#8c8c94' #gray
 69
 70
     c2 = '#ffd56f' #pale yellow
     c3 = '#005e7b' #sea blue
     c4 = '#a06534' #reddish brown
 72
 73
     c5 = '#404436' #rifle green
 74
     c6 = '#7a26e7' #magenta
 75
     c7 = '#5CCE2A' #pale green
 76
     c8 = '#000000' #black
     c9 = '#4542f4' #purple
 77
 78
     #Put all of the colors into a global colors list.
 79
     c = [c0, c1, c2, c3, c4, c5, c6, c7, c8, c9]
 80
 81
     #What follows are two global initial conditions arrays. orbit() looks to this
 82
     #data, but only imports the appropriate amount of rows for the given N.
 83
     \#For\ r0 , the first row of data represents m0's (x,y,z) initial position, the
 84
 85
     #second row represents m1's (x,y,z) initial position, etc.
     r0_{=} np.array([[ 1.0, 3.0, 2.0 ], #0]
 86
                    [ 6.0, -5.0, 4.0], #1
 87
 88
                      7.0, 8.0, -7.0 ], #2
                    [
 89
                    [ 8.0, 6.0, -2.0], #3
                    [ 8.8, 9.8, -6.8], #4
 90
                    [ 9.8, 3.8, -7.8], #5
 91
                    [ -3.8, 1.8, 4.8], #6
[ 7.8, -2.2, 1.8], #7
 92
 93
 94
                      6.8, -4.1, 3.8 ], #8
                    [
 95
                    [ 5.8, -9.3, 5.8]])*1e11
 96
     \#For\ v0 , the first row of data represents m0's (x,y,z) initial velocity, the
     #second row represents m1's (x,y,z) initial velocity, etc.
 97
 98
     v0 = np.array([[ 0.0, 0.0, 0.0], #0]
                    [ 7.0, 0.5, 2.0], #1
99
                    [ -4.0, -0.5, -3.0 ], #2
100
101
                    [ 7.0, 0.5, 2.0 ], #3
102
                    [ 4.8, 1.3, 4.8], #4
103
                    [ 1.8, 1.2, -5.8], #5
                    [ 2.8, 11.3, 1.4], #6
104
105
                      3.8, 10.3, 2.4], #7
                    [
                      4.8, 9.3, -1.4], #8
106
                    [
                      5.8, 0.3, -2.4 ]]) *1e3
107
108
     109
     110
111
     11 11 11
112
113
    Purpose:
114
         orbit() calculates the orbital trajectories of N gravitationally
115
         interacting bodies given a set of mass and initial conditions data. Note
116
         both the mass list and initial conditions arrays must each contain data for
117
         at least N bodies, but may contain more. For example, orbit() could plot
118
         trajectories for the first 5 of 100 bodies in a large database.
119
     Inputs:
120
         [0] N = the number of bodies to be considered in the calculation (integer)
121
         [1] t0 = the start time in years (number)
122
         [2] tf = the end time in years (number)
123
         [3] dt = the time step in hours (number)
124
         [4] m = list of masses of (at least) length N (list of numbers)
125
         [5] r0 = (at least) an Nx3 array of initial position data (2D numpy array)
126
         [6] v0 = (at least) an Nx3 array of initial velocity data (2D numpy array)
127
         [7] method = choice of stepping method (string)
128
129
    Outputs:
         [0] r = position data (3D numpy array)
130
131
         [1] v = velocity data (3D numpy array)
132
         [2] p = momentum data (3D numpy array)
133
         [3] KE = kinetic energy data (3D numpy array)
134
         [4] T = total kinetic energy data (3D numpy array)
```

```
136
137
      def orbit(N, t0, tf, dt, m, r0, v0, method):
138
          G = 6.67e-11 #universal gravitational constant in units of m<sup>3</sup>/kg*s<sup>2</sup>
139
          t0 = t0*365.26*24*3600 \# Convert t0 from years to seconds.
140
          tf = tf*365.26*24*3600 \# Convert tf from years to seconds.
141
          dt = dt*3600.0 #Convert dt from hours to seconds.
142
          steps = int(abs(tf-t0)/dt) #Define the total number of time steps.
          #Multiply an array of integers [(0, 1, ..., steps-1, steps)] by dt to get
143
144
          #an array of ascending time values.
145
          t = dt*np.array(range(steps + 1))
146
147
          #If you print out either r or v below, you'll see several "layers" of Nx3
148
          #matrices. Which layer you are on represents which time step you are on.
149
          #Within each Nx3 matrix, the row denotes which body, while the columns 1-3
150
          #(indexed 0-2 in Python) represent x-, y-, z-positions respectively. In
151
          #essence, each number in r or v is associated with three indices:
152
          #step #, body #, and coordinate #.
          r = np.zeros([steps+1, N, 3])
153
154
          v = np.zeros([steps+1, N, 3])
155
          r[0] = r0[0:N] #Trim the initial conditions arrays to represent N bodies.
156
          v[0] = v0[0:N]
157
          11 11 11
158
159
          Purpose:
160
              accel() acts as a subroutine for orbit() by returning 3D acceleration
161
              vectors for a number of gravitationally interacting bodies given each of
162
              their positions.
163
164
              [0] r = 3D position vectors for all bodies at a certain time step
165
                      (Nx3 numpy array)
166
          Output:
167
              [0] a = 3D acceleration vectors for each body (Nx3 numpy array)
168
169
          def accel(r):
170
              a=np.zeros([N,3])
              #Each body's acceleration at each time step has to do with forces from
171
172
              #all other bodies. See: https://en.wikipedia.org/wiki/N-body problem
173
              for i in range(N):
                  #j is a list of indices of all bodies other than the ith body.
174
175
                  j = list(range(i)) + list(range(i+1,N))
176
                  #Note each body's acceleration vector is a sum of N-1 terms, so for
177
                  #each body, the terms are successively added to each other in a
178
                  #running sum. Once all terms are added together, the ith body's
179
                  #acceleration vector results.
180
                  for k in range(N-1):
181
                      a[i]=G*(m[j[k]]/LA.norm(r[i]-r[j[k]])**3*(r[j[k]]-r[i]))+a[i]
182
              return a
183
184
          #The simplest way to numerically integrate the accelerations into
185
          #velocities and then positions is with the Euler method. Note that this
186
          #method does not conserve energy.
187
          if method == 'euler':
188
              for i in range(steps):
189
                  r[i+1] = r[i] + dt*v[i]
190
                  v[i+1] = v[i] + dt*accel(r[i])
191
192
          #The Euler-Cromer method drives our next-simplest stepper.
193
          if method == 'ec':
194
              for i in range(steps):
195
                  r[i+1] = r[i] + dt*v[i]
196
                  v[i+1] = v[i] + dt*accel(r[i+1])
197
198
          #Getting slightly fancier, we employ the 2nd Order Runge-Kutta method.
199
          if method == 'rk2':
200
              for i in range(steps):
201
                  v iphalf = v[i] + accel(r[i])*(dt/2) # (i.e. v[i+0.5])
```

135

[5] t = time data (1D numpy array)

```
r_{iphalf} = r[i] + v[i]*(dt/2)
203
                v[i+1] = v[i] + accel(r iphalf)*dt
204
                r[i+1] = r[i] + v iphalf*dt
205
206
         #Here is a velocity Verlet implementation.
207
         #See: http://young.physics.ucsc.edu/115/leapfrog.pdf
208
         if method == 'vv':
209
             for i in range(steps):
210
                v iphalf = v[i] + (dt/2)*accel(r[i])
211
                r[i+1] = r[i] + dt*v iphalf
212
                v[i+1] = v iphalf + (dt/2)*accel(r[i+1])
213
214
         #Next is a position Verlet implementation (found in the same pdf as 'vv').
215
         if method == 'pv':
216
             for i in range(steps):
217
                r iphalf = r[i] + (dt/2)*v[i]
218
                v[i+1] = v[i] + dt*accel(r iphalf)
219
                r[i+1] = r iphalf + (dt/2)*v[i+1]
220
221
         #EFRL refers to an extended Forest-Ruth-like integration algorithm. Below
222
         #are three optimization parameters associated with EFRL routines.
223
         e = 0.1786178958448091e0
224
         1 = -0.2123418310626054e0
225
         k = -0.6626458266981849e-1
226
         #First we do a velocity EFRL implementation (VEFRL).
227
         #See: https://arxiv.org/pdf/cond-mat/0110585.pdf
228
         if method == 'vefrl':
229
             for i in range(steps):
230
                v1 = v[i] + accel(r[i])*e*dt
231
                r1 = r[i] + v1*(1-2*1)*(dt/2)
232
                v2 = v1 + accel(r1)*k*dt
233
                r2 = r1 + v2*1*dt
234
                v3 = v2 + accel(r2)*(1-2*(k+e))*dt
235
                r3 = r2 + v3*1*dt
236
                v4 = v3 + accel(r3)*k*dt
237
                r[i+1] = r3 + v4*(1-2*1)*(dt/2)
                v[i+1] = v4 + accel(r[i+1])*e*dt
238
239
240
         #Next is a position EFRL (PEFRL) (found in the same pdf as 'vefrl').
         if method == 'pefrl':
241
242
             for i in range(steps):
2.43
                r1 = r[i] + v[i] *e*dt
244
                v1 = v[i] + accel(r1)*(1-2*1)*(dt/2)
245
                r2 = r1 + v1*k*dt
246
                v2 = v1 + accel(r2)*l*dt
                r3 = r2 + v2*(1-2*(k+e))*dt
247
248
                v3 = v2 + accel(r3)*l*dt
249
                r4 = r3 + v3*k*dt
250
                v[i+1] = v3 + accel(r4)*(1-2*1)*(dt/2)
251
                r[i+1] = r4 + v[i+1] *e*dt
252
253
         #Lastly, derive kinetic energy and momentum data from the velocity data.
254
         T = np.zeros((steps+1,N,1)) #array for total kinetic energy data
255
         KE = np.zeros((steps+1,N,3)) #array for 3D kinetic energy data
256
         p = np.zeros((steps+1,N,3)) #array for 3D momentum data
257
         v2 = v**2 #Square all velocities for use in the energy calculation.
258
         for i in range(steps):
259
             for j in range(N):
260
                KE[i,j,:] = (m[j]/2)*v2[i,j,:]
261
                T[i,j,0] = sum(KE[i,j,:])
262
                p[i,j,:] = m[j] * v[i,j,:]
263
264
         return r, v, p, KE, T, t
265
266
     267
     268
```

```
269
     r, v, p, KE, T, t = orbit(N , t0 , tf , dt , m , r0 , v0 , method)
270
271
     272
     273
     #Select a body index from 0, 1, ..., N-1. Phase space and energy plots will be
274
275
     #generated for the selected body.
276
277
     mf lab = '$m '+str(mf)+'$' #Create a LaTeX-wrapped label for mf.
278
279
     #Generate an ascending list of integers from 2 to N and then change the
280
     #elements into strings for use in figure titles.
281
     words = list(range(^2,N +^1))
282
     words = [str(x) for x in words]
283
     #Wrap each string in LaTeX so it has a serif font on the plot.
284
     for i in range(len(words)):
285
         words[i] = '$\mathrm{'+words[i]+'\ }$'
286
     Nstr = words[N -2] #Note the index shift because words[0]='2'.
287
288
     labs = [None] *N  #Create a list to store labels for the masses (m0, m1, etc.).
     #Wrap each label with LaTeX math mode so it prints with a serif font.
289
     for i in range(N ):
290
291
         labs[i] = r'$m '+str(i)+'$'
292
293
     a_{-} = 0.7 #Set a global transparency value so we can see where orbits overlap.
294
     #-----
295
     fig1 = plt.figure(1, facecolor='white') #3D plot of orbital trajectories
296
     ax1 = fig1.add subplot(1,1,1, projection='3d')
297
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies}$', y=1.05)
     ax1.set xlabel(r'$\mathrm{x-Position}\ \mathrm{(m)}$', labelpad=10)
298
     ax1.set_ylabel(r'$\mathrm{y-Position}\ \mathrm{(m)}$', labelpad=10)
299
     ax1.set zlabel(r'$\mathrm{z-Position}\ \mathrm{(m)}$', labelpad=10)
300
301
     for i in range(N): \#For all times, plot mi's (x,y,z) data.
302
         ax1.plot(r[:, i, 0], r[:, i, 1], r[:, i, 2], color=c [i], label=labs[i],
303
             alpha=a )
304
     ax1.axis('equal')
305
     plt.legend(loc='upper left')
306
     plt.savefig('fig1', bbox inches='tight')
     #-----
307
     fig2 = plt.figure(2, facecolor='white') #fig1 as viewed from the +x-axis
308
     ax2 = fig2.add subplot (111)
309
310
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
311
              +r'$\mathrm{as\ Viewed \ From\ the\ Positive\ x-Axis}$', y=1.05)
312
     ax2.set xlabel(r'$\mathrm{y-Position}\ \mathrm{(m)}$')
     ax2.set ylabel(r'$\mathrm{z-Position}\ \mathrm{(m)}$')
313
314
     for i in range(N ): #For all times, plot mi's (y,z) data.
315
         ax2.plot(r[:, i, 1], r[:, i, 2], color=c [i], label=labs[i], alpha=a )
316
     ax2.axis('equal')
317
     ax2.legend(loc='lower right')
318
     plt.savefig('fig2', bbox inches='tight')
319
320
     fig3 = plt.figure(3, facecolor='white') #fig1 as viewed from the +y-axis
321
     ax3 = fig3.add subplot (111)
322
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
323
              +r'$\mathrm{as\ Viewed \ From\ the\ Positive\ y-Axis}$', y=1.05)
     ax3.set_xlabel(r'$\mathrm{x-Position}\ \mathrm{(m)}$')
324
325
     ax3.set ylabel(r'$\mathrm{z-Position}\ \mathrm{(m)}$')
326
     for i in range(N ): #For all times, plot mi's (x,z) data.
327
         ax3.plot(r[:, i, 0], r[:, i, 2], color=c[i], label=labs[i], alpha=a)
328
     ax3.axis('equal')
329
     ax3.legend(loc='lower right')
330
     plt.savefig('fig3', bbox inches='tight')
331
     fig4 = plt.figure(4, facecolor='white') #fig1 as viewed from the +z-axis
332
333
     ax4 = fig4.add subplot(111)
334
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
335
              +r'$\mathrm{as\ Viewed \ From\ the\ Positive\ z-Axis}$', y=1.05)
```

```
336
     ax4.set xlabel(r'$\mathrm{x-Position}\ \mathrm{(m)}$')
     ax4.set_ylabel(r'$\mathrm{y-Position}\ \mathrm{(m)}$')
337
     for i in range(N ): #For all times, plot mi's (x,y) data.
338
339
         ax4.plot(r[:, i, 0], r[:, i, 1], color=c_[i], label=labs[i], alpha=a_)
340
     ax4.axis('equal')
341
     ax4.legend(loc='lower right')
342
     plt.savefig('fig4', bbox inches='tight')
343
     #-----
344
     #x-Component Momentum Phase Space for mf
345
346
     fig5 = plt.figure(5, facecolor='white')
347
     ax5 = fig5.add subplot (111)
348
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
349
              +r'$\mathrm{x-Component\ Momentum\ Phase\ Space\ for\ }$'
350
              +r'%s'%mf lab, y=1.05)
351
     ax5.set xlabel(r'$\mathrm{x-Position}\ \mathrm{(m)}$')
     ax5.set ylabel(r'$\mathrm{x-Momentum}\ \mathrm{(kg\cdot\frac{m}{s}))}$')
352
353
     ax5.plot(r[:, mf, 0], p[:, mf, 0], color=c [mf], label=labs[mf], alpha=a)
     plt.savefig('fig5', bbox inches='tight')
354
355
     #-----
356
     #y-Component Momentum Phase Space for mf
357
     fig6 = plt.figure(6, facecolor='white')
358
     ax6 = fig6.add subplot(111)
359
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
360
              +r'$\mathrm{y-Component\ Momentum\ Phase\ Space\ for\ }$'
361
              +r'%s'%mf_lab, y=1.05)
362
     ax6.set xlabel(r'$\mathrm{y-Position}\ \mathrm{(m)}$')
363
     ax6.set ylabel(r'$\mathrm{y-Momentum}\ \mathrm{(kg\cdot\frac{m}{s}))}$')
364
     ax6.plot(r[:, mf, 1], p[:, mf, 1], color=c [mf], label=labs[mf], alpha=a)
365
     plt.savefig('fig6', bbox_inches='tight')
366
     #-----
367
     #z-Component Momentum Phase Space for mf
368
     fig7 = plt.figure(7, facecolor='white')
369
     ax7 = fig7.add subplot (111)
370
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
              +r'$\mathrm{z-Component\ Momentum\ Phase\ Space\ for\ }$'
371
              +r'%s'%mf lab, y=1.05)
372
373
     ax7.set xlabel(r'$\mathrm{z-Position}\ \mathrm{(m)}$')
     ax7.set_ylabel(r'$\mathrm{z-Momentum}\ \mathrm{(kg\cdot\frac{m}{s}))}$')
374
375
     ax7.plot(r[:, mf, 2], p[:, mf, 2], color=c [mf], label=labs[mf], alpha=a)
376
     plt.savefig('fig7', bbox inches='tight')
377
     #-----
378
379
     #x-Component Kinetic Energy Phase Space for mf
380
     fig8 = plt.figure(8, facecolor='white')
381
     ax8 = fig8.add subplot(111)
382
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies}$'+'\n'
              +r'$\mathrm{x-Component\ Kinetic\ Energy\ Phase\ Space\ for\ }$'
383
384
              +r'%s'%mf lab, y=1.05)
385
     ax8.set xlabel(r'$\mathrm{x-Position}\ \mathrm{(m)}$')
386
     ax8.set ylabel(r'$\mathrm{x-Kinetic\ Energy}\ \mathrm{(J)}$')
     ax8.plot(r[:, mf, 0], KE[:, mf, 0], color=c [mf], label=labs[mf], alpha=a )
387
     plt.savefig('fig8', bbox_inches='tight')
388
     #-----
389
390
     #y-Component Kinetic Energy Phase Space for mf
391
     fig9 = plt.figure(9, facecolor='white')
392
     ax9 = fig9.add subplot (111)
393
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies}$'+'\n'
394
              +r'$\mathrm{y-Component\ Kinetic\ Energy\ Phase\ Space\ for\ }$'
395
              +r'%s'%mf lab, y=1.05)
     ax9.set xlabel(r'$\mathrm{y-Position}\ \mathrm{(m)}$')
396
397
     ax9.set ylabel(r'$\mathrm{y-Kinetic\ Energy}\ \mathrm{(J)}$')
398
     ax9.plot(r[:, mf, 1], KE[:, mf, 1], color=c [mf], label=labs[mf], alpha=a )
     plt.savefig('fig9', bbox_inches='tight')
399
400
     #-----
401
     #z-Component Kinetic Energy Phase Space for mf
402
     fig10 = plt.figure(10, facecolor='white')
```

```
403
     ax10 = fig10.add subplot(111)
404
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies}$'+'\n'
405
                +r'$\mathrm{z-Component\ Kinetic\ Energy\ Phase\ Space\ for\ }$'
406
                +r'%s'%mf_lab, y=1.05)
     ax10.set xlabel(r'$\mathrm{z-Position}\ \mathrm{(m)}$')
407
408
      ax10.set ylabel(r'$\mathrm{z-Kinetic\ Energy}\ \mathrm{(J)}$')
409
      ax10.plot(r[:, mf, 2], KE[:, mf, 2], color=c [mf], label=labs[mf], alpha=a )
410
      plt.savefig('fig10', bbox_inches='tight')
411
412
      #Total Kinetic Energy over Time for All Bodies
413
      fig11 = plt.figure(11, facecolor='white')
414
415
      ax11 = fig11.add subplot(111)
416
     plt.title(r'%s'%Nstr+r'$\mathrm{Orbiting\ Bodies\ }$'+'\n'
                +r'$\mathrm{Total\ Kinetic\ Energy\ vs.\ Time}$', y=1.05)
417
      ax11.set xlabel(r'$\mathrm{Time}\ \mathrm{(s)}$')
418
      ax11.set ylabel(r'$\mathrm{Total\ Kinetic\ Energy}\ \mathrm{(J)}$')
419
420
     for i in range(N ):
         ax11.plot(t, T[:, i, :], color=c_[i], label=labs[i], alpha=a )
421
422
     ax11.legend(loc='lower right')
423
     plt.savefig('fig11', bbox inches='tight')
```

424