ode

August 5, 2021

1 The Stability of the Planetary Orbits

1.1 Call the C library ODE solver

```
[1]: import ctypes
     from ctypes import *
     from numpy.ctypeslib import ndpointer
     import numpy as np
     import matplotlib.pyplot as plt
     import numba
     import csv
     import random
     # ODE solver
     import odesolver
     from odesolver import *
     # neural network model
     import model
     from model import *
     # equations of motion, transformation of coordinates
     import physics
     from physics import *
```

```
mpl.rcParams['legend.fontsize'] = 12
mpl.rcParams['legend.handlelength'] = 2
mpl.rcParams['axes.titlesize'] = 20
mpl.rcParams['xtick.labelsize'] = 20
mpl.rcParams['ytick.labelsize'] = 20
mpl.rcParams['axes.labelsize'] = 15
mpl.rcParams['mathtext.fontset'] = 'cm'
mpl.rcParams['axes.unicode_minus']=False
mpl.rcParams['figure.figsize'] = (8, 4)
mpl.rcParams['figure.dpi'] = 300
mpl.rcParams['agg.path.chunksize'] = 10000
```

```
[1]: #!gcc -shared -02 ode.c -o libode.so
```

- 1.2 The right-hand side of the equation of motion: $\frac{d\vec{z}}{dt} = \vec{f}(t, \vec{z})$ is in physics.py
- 1.2.1 Mercury, Venus, Earth, Mars, Jupiter, Saturn, and the Sun
- 1.3 Run the simulation
- 1.3.1 Initial conditions:

```
Mercury a_m = 0.39 AU, e_m = 0.206, \theta_E = -3\pi/4, \theta_m = \theta_E
```

Venus $M_V = 0.000002447 M_S$, $a_V = 0.72 AU$, $T_V = 0.615 years$, $\theta_V = 0$

Earth $M_E = M_S/333030$, $a_E = 1AU$, $T_E = 1 years$, $\theta_E = 0$

Mars $M_M = 0.0000003213 M_S$, $a_M = 1.52 AU$, $T_M = 1.88 years$, $\theta_M = 0$

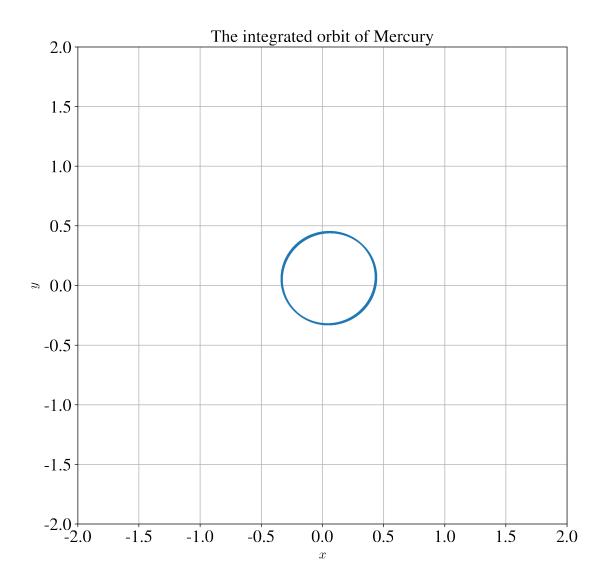
Jupiter $M_J = M_S/1048$, $a_J = 5.20 AU$, $T_J = 11.86 years$, $\theta_J = 0$

Saturn $M_S = 0.0002857 M_S$, $a_S = 9.58 AU$, $T_J = 29.46 years$, $\theta_J = 0$

```
[6]: # run the simulation
      dt = 0.001
      time_f = 10000.
      nsteps = int(time_f/dt)
      t,zm = solve_ode(funcpl,[0.,time_f], nsteps, init_cart, args=planets,__
      →method="Yoshida4")
      xm, Vmx, ym, Vmy = zm.T
      # compute the parameters of the ellipse
      a,e,theta,thetaE = xy_to_ellipse(xm,Vmx,ym,Vmy)
[36]: plt.figure(figsize=(10,10))
      plt.title(r"The integrated orbit of Mercury")
      plt.ylabel(r"$y$")
      plt.xlabel(r"$x$")
      plt.plot(xm,ym)
      #plt.legend()
      plt.grid()
```

plt.savefig("mercury_planets.png",bbox_inches="tight")

axes = plt.gca()
axes.set_xlim([-2,2])
axes.set_ylim([-2,2])



1.4 Labeling the data

Decide whether Mercury is still on the orbit

```
[38]: # if major semiaxis exceeds 10 AU, Mercury is ejected from the system
# if stability is 1, Mercury is not ejected; otherwise, we say that it is
→ ejected

stability = int(np.count_nonzero((a>10))==0)
print(stability)
```

1

1.5 Running simulations with pertubations

Modify the initial conditions of Mercury

```
thetaEm = -0.75 * np.pi
      thetam = thetaEm
      init_cart = np.array(ellipse_to_xy(am,em,thetaEm))
      planets = np.array([0.000002447, 0.72, 0.615,
                          1./333030., 1., 1.,
                          0.0000003213, 1.52, 1.88,
                          1./1048., 5.20, 11.86,
                          0.0002857, 9.58, 29.46])
      dt = 0.001
      time_f = 10000.
      nsteps = int(time_f/dt)
      with open('train.csv', mode='a', newline='') as file:
         for iteration in range(0,500):
              # add pertubations to the orbit of Mercury
              am = random.uniform(0.1,5.0)
              em = random.gauss(0.206, 0.01)
              init_cart = np.array(ellipse_to_xy(am,em,thetaEm))
              # run the simulation
              t,zm = solve_ode(funcpl,[0.,time_f], nsteps, init_cart, args=planets,_u
      →method="Yoshida4")
             xm,Vmx,ym,Vmy = zm.T
              # compute the parameters of the ellipse
              a,e,theta,thetaE = xy_to_ellipse(xm,Vmx,ym,Vmy)
              # 1 if stable, 0 if unstable
              stability = int(np.count_nonzero((a>10))==0)
              writer = csv.writer(file, delimiter=',', quotechar='"', quoting=csv.
      →QUOTE_MINIMAL)
              data = [am, em, stability]
              writer.writerow(data)
[33]: # test set
      thetaEm = -0.75 * np.pi
      thetam = thetaEm
      init_cart = np.array(ellipse_to_xy(am,em,thetaEm))
      planets = np.array([0.000002447, 0.72, 0.615,
                          1./333030., 1., 1.,
                          0.0000003213, 1.52, 1.88,
```

[18]: # training set

```
1./1048., 5.20, 11.86,
                    0.0002857, 9.58, 29.46])
dt = 0.001
time_f = 10000.
nsteps = int(time_f/dt)
with open('test.csv', mode='a', newline='') as file:
   for iteration in range (0,250):
        # add pertubations to the orbit of Mercury
       am = random.uniform(0.1,5.0)
        em = random.gauss(0.206, 0.01)
        init_cart = np.array(ellipse_to_xy(am,em,thetaEm))
        # run the simulation
       t,zm = solve_ode(funcpl,[0.,time_f], nsteps, init_cart, args=planets,__
→method="Yoshida4")
       xm,Vmx,ym,Vmy = zm.T
        # compute the parameters of the ellipse
        a,e,theta,thetaE = xy_to_ellipse(xm,Vmx,ym,Vmy)
        # 1 if stable, 0 if unstable
        stability = int(np.count_nonzero((a>10))==0)
       writer = csv.writer(file, delimiter=',', quotechar='"', quoting=csv.
 →QUOTE_MINIMAL)
       data = [am, em, stability]
        writer.writerow(data)
```

1.6 Predicting the stability of the system

```
print(Y_train.shape)
      # reshape X_train and Y_train to avoid having rank 1 array
      X_train = X_train.reshape(2,m_train)
      Y_train = Y_train.reshape(1,m_train)
      print(X_train.shape)
      print(Y_train.shape)
     500
     (2, 500)
     (500,)
     (2, 500)
     (1, 500)
[26]: # Build a model with a n h-dimensional hidden layer
      parameters = nn_model(X_train, Y_train, n_h=4, num_iterations=10000, u
       →print_cost=True)
      # Print accuracy
      predictions_train = predict(parameters, X_train)
     Cost after iteration 0: 0.693035
     Cost after iteration 1000: 0.288190
     Cost after iteration 2000: 0.272064
     Cost after iteration 3000: 0.253229
     Cost after iteration 4000: 0.249525
     Cost after iteration 5000: 0.247474
     Cost after iteration 6000: 0.246544
     Cost after iteration 7000: 0.246056
     Cost after iteration 8000: 0.245648
     Cost after iteration 9000: 0.245290
[27]: accuracy_train = (np.count_nonzero(predictions_train[0] == Y_train[0]))/m_train_
       →* 100.
      print("Accuracy for the training set: %0.1f %%" %accuracy_train)
     Accuracy for the training set: 90.8 %
[28]: # read data
      a_test,e_test,stability_test = np.loadtxt("test.csv",delimiter=',',unpack=True)
      m_test = a_test.shape[0]
      print(m_test)
      # create an array of X and Y
      X test = np.zeros((2,m test))
      X_test[0,:] = a_test
      X \text{ test}[1,:] = e \text{ test}
```

```
Y_test = stability_test
      print(X_test.shape)
      print(Y_test.shape)
      # reshape X_test and Y_test
      X_test = X_test.reshape(2,m_test)
      Y_test = Y_test.reshape(1,m_test)
      print(X_test.shape)
     print(Y_test.shape)
     250
     (2, 250)
     (250,)
     (2, 250)
     (1, 250)
[29]: # Compute and print accuracy
     predictions = predict(parameters, X_test)
      accuracy_test = (np.count_nonzero(predictions[0] == Y_test[0]))/m_test * 100.
     print("Accuracy for the training set: %0.1f %%" %accuracy_test)
```

Accuracy for the training set: 91.6 %