

Double Pendulum and Multiple Pendulum

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1 Description

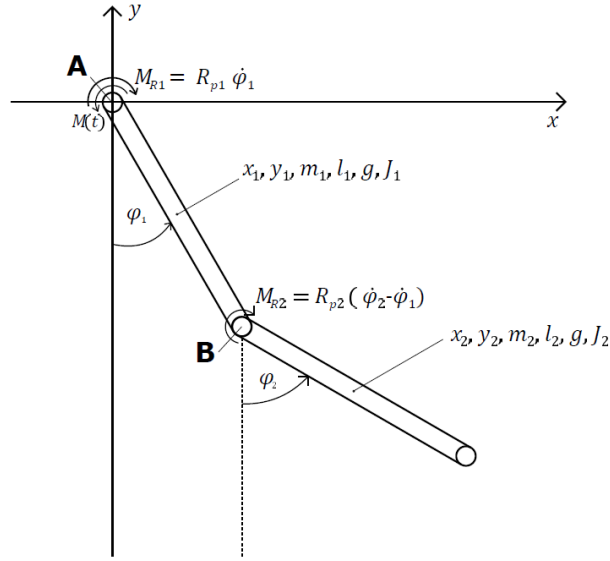


Figure 1: Schematic of a double pendulum

A double pendulum is a combination of two simple pendulums, one attached to the end of another.

We describe the system using R_0, R_1 , the lengths of two pendulums, m_0, m_1 , the masses of two ends, φ_0, φ_1 , the angles of the rods.

We will derive the dynamical equations and simulate the dynamics of the system.

2 Dynmanics

We write down the kinetic energy and potential of the system,

$$\begin{aligned}
K &= \frac{1}{2}m_0 (R_0\dot{\varphi}_0)^2 + \frac{1}{2}m_1 ((R_0\dot{\varphi}_0 \cos \varphi_0 + R_1\dot{\varphi}_1 \cos \varphi_1)^2 + (R_0\dot{\varphi}_0 \sin \varphi_0 + R_1\dot{\varphi}_1 \sin \varphi_1)^2) \\
&= \frac{1}{2}m_0 (R_0\dot{\varphi}_0)^2 + \frac{1}{2}m_1 ((R_0\dot{\varphi}_0)^2 + (R_1\dot{\varphi}_1)^2 + 2R_0R_1\dot{\varphi}_0\dot{\varphi}_1 \cos(\varphi_0 - \varphi_1)) \\
V &= -gm_0R_0 \cos \varphi_0 - gm_1R_0 \cos \varphi_0 - gm_1R_1 \cos \varphi_1 \\
\mathcal{L} &= K - V
\end{aligned}$$

Compute each partial derivatives of Lagrangian,

$$\begin{aligned}
\frac{\partial \mathcal{L}}{\partial \dot{\varphi}_0} &= \frac{\partial K}{\partial \dot{\varphi}_0} = m_0R_0^2\dot{\varphi}_0 + m_1 (R_0^2\dot{\varphi}_0 + R_0R_1\dot{\varphi}_1 \cos(\varphi_0 - \varphi_1)) \\
\frac{\partial \mathcal{L}}{\partial \dot{\varphi}_1} &= \frac{\partial K}{\partial \dot{\varphi}_1} = m_1 (R_1^2\dot{\varphi}_1 + R_0R_1\dot{\varphi}_0 \cos(\varphi_0 - \varphi_1)) \\
\frac{\partial \mathcal{L}}{\partial \varphi_0} &= -m_1R_0R_1\dot{\varphi}_0\dot{\varphi}_1 \sin(\varphi_0 - \varphi_1) - gm_0R_0 \sin \varphi_0 - gm_1R_0 \sin \varphi_0 \\
\frac{\partial \mathcal{L}}{\partial \varphi_1} &= m_1R_0R_1\dot{\varphi}_0\dot{\varphi}_1 \sin(\varphi_0 - \varphi_1) - gm_1R_1 \sin \varphi_1 \\
\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\varphi}_0} &= m_0R_0^2\ddot{\varphi}_0 + m_1 (R_0^2\ddot{\varphi}_0 + R_0R_1\ddot{\varphi}_1 \cos(\varphi_0 - \varphi_1) - R_0R_1\dot{\varphi}_1 (\dot{\varphi}_0 - \dot{\varphi}_1) \sin(\varphi_0 - \varphi_1)) \\
\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\varphi}_1} &= m_1 (R_1^2\ddot{\varphi}_1 + R_0R_1\ddot{\varphi}_0 \cos(\varphi_0 - \varphi_1) - R_0R_1\dot{\varphi}_0 (\dot{\varphi}_0 - \dot{\varphi}_1) \sin(\varphi_0 - \varphi_1))
\end{aligned}$$

Consider Euler-Lagrange equation,

$$\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\varphi}_k} - \frac{\partial \mathcal{L}}{\partial \varphi_k} = 0$$

$$\begin{aligned}
(m_0 + m_1)R_0\ddot{\varphi}_0 + m_1R_1 \cos(\varphi_0 - \varphi_1) \ddot{\varphi}_1 &= -m_1R_1\dot{\varphi}_1^2 \sin(\varphi_0 - \varphi_1) - g(m_0 + m_1) \sin \varphi_0 \\
m_1R_0 \cos(\varphi_0 - \varphi_1) \ddot{\varphi}_0 + m_1R_1\ddot{\varphi}_1 &= m_1R_0\dot{\varphi}_0^2 \sin(\varphi_0 - \varphi_1) - gm_1 \sin \varphi_1
\end{aligned}$$

Since $\varphi_0, \varphi_1, \dot{\varphi}_0, \dot{\varphi}_1$ are known, this is a linear equation to $\ddot{\varphi}_0, \ddot{\varphi}_1$,

$$\begin{aligned}
a\ddot{\varphi}_0 + b\ddot{\varphi}_1 &= e \\
c\ddot{\varphi}_0 + d\ddot{\varphi}_1 &= f
\end{aligned}$$

where

$$\begin{aligned}
a &= (m_0 + m_1)R_0 \\
b &= m_1R_1 \cos(\varphi_0 - \varphi_1) \\
c &= m_1R_0 \cos(\varphi_0 - \varphi_1) \\
d &= m_1R_1 \\
e &= -m_1R_1\dot{\varphi}_1^2 \sin(\varphi_0 - \varphi_1) - g(m_0 + m_1) \sin \varphi_0 \\
f &= m_1R_0\dot{\varphi}_0^2 \sin(\varphi_0 - \varphi_1) - gm_1 \sin \varphi_1
\end{aligned}$$

The solution is,

$$\ddot{\varphi}_0 = \frac{de - bf}{ad - bc}$$

$$\ddot{\varphi}_1 = \frac{af - ce}{ad - bc}$$

3 Program

The program is straightforward. We define a class called double pendulum and define functions to compute Lagrangian and solution as shown above.

```

1 # xi = [phi_0, phi_1, phi_dot_0, phi_dot_1]
2 g = 9.8
3
4 class double_pendulum:
5     def __init__(self, R, M, Phi0, ode = rk4) -> None:
6         self.R = R # length
7         self.M = M # mass
8         self.Phi0 = Phi0 # initial angles
9         self.xi0 = np.array([Phi0[0], Phi0[1], 0, 0])
10        self.ode = ode # need to appoint a ode solver
11        self.step = 0.01 # default step
12
13    def diff_xi(self, xi, t): # compute d(xi)/dt
14        diff_xi = np.array([xi[2], xi[3], 0, 0])
15        phi = np.array([xi[0], xi[1]])
16        phi_dot = np.array([xi[2], xi[3]])
17        m = self.M
18        R = self.R
19
20        a = (m[0]+m[1])*(R[0]**2)
21        b = c = m[1]*R[0]*R[1]*np.cos(phi[0]-phi[1])
22        d = m[1]*(R[1]**2)
23        e = -m[1]*R[0]*R[1]*(phi_dot[1]**2)*np.sin(phi[0]-phi[1])-g*(m[0]+m[1])*R[0]*np
            .sin(phi[0])
24        f = m[1]*R[0]*R[1]*(phi_dot[0]**2)*np.sin(phi[0]-phi[1])-g*m[1]*R[1]*np.sin(phi
            [1])
25
26        D = a*d - b*c
27        diff_xi[2] = (d*e - b*f) / D
28        diff_xi[3] = (a*f - c*e) / D
29        return diff_xi
30
31    def Ek(self, xi): # kinetic energy
32        phi = np.array([xi[0], xi[1]])
33        phi_dot = np.array([xi[2], xi[3]])
34        m = self.M

```

```

35     R = self.R
36     return m[0]/2*(R[0]*phi_dot[0])**2 + m[1]/2*((R[0]*phi_dot[0])**2 + (R[1]*
37         phi_dot[1])**2 \
38         + 2*R[0]*R[1]*phi_dot[0]*phi_dot[1]*np.cos(phi[0] - phi[1]))
39
40 def V(self, xi): # potential
41     phi = np.array([xi[0], xi[1]])
42     m = self.M
43     R = self.R
44     return -g*(m[0]*R[0]*np.cos(phi[0]) + m[1]*R[0]*np.cos(phi[0]) + m[1]*R[1]*np.
45         cos(phi[1]))
46
47 def Lagrangian(self, xi):
48     return self.Ek(xi) - self.V(xi)
49
50 def Hamiltonian(self, xi):
51     return self.Ek(xi) + self.V(xi)
52
53 def simulate(self, t0, t1, h = 0.1):
54     T, Xi = self.ode(self.diff_xi, t0, self.xi0, t1, h)
55     return T, Xi

```

We write two ODE solvers for comparison.

```

1  # f: return dx/dt, t0: start time, x0: initial condition, ti: end time, h: step
2  def rk4(f, t0, x0, ti, h = 1): # 4th-order Runge-Kutta
3      T = [t0]
4      X = [x0]
5      while T[-1] < ti:
6          x, t = X[-1], T[-1]
7          k1 = h * f(x, t)
8          k2 = h * f(x + 0.5*k1, t + 0.5*h)
9          k3 = h * f(x + 0.5*k2, t + 0.5*h)
10         k4 = h * f(x + k3, t + h)
11         T.append(t + h)
12         X.append(x + (k1 + 2*k2 + 2*k3 + k4)/6)
13     return np.array(T), np.array(X)
14
15 def leapfrog(f, t0, x0, ti, h = 1):
16     T = [t0]
17     X = [x0]
18     x_half = x0 + 0.5 * h * f(x0, t0)
19     while T[-1] < ti:
20         x, t = X[-1], T[-1]
21         T.append(t + h)
22         X.append(x + h * f(x_half, t + 0.5*h))
23         x_half += h * f(X[-1], T[-1])
24     return np.array(T), np.array(X)

```

As for plot and animation parts, please refer to the source codes.

4 Performance

We plot total energy using rk4 and leapfrog with smaller steps to see whether they converge.

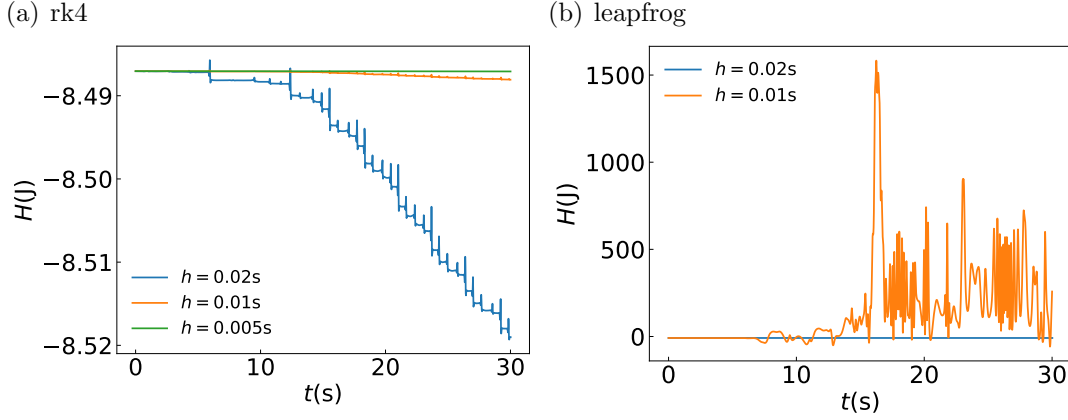


Figure 2: Comparison of convergence between rk4 and leapfrog algorithm. The plot is Hamiltonian versus time. Total energy using rk4 converges but that of leapfrog method does not converge. Initial condition is $R_0 = 2, R_1 = 1, m_0 = 2, m_1 = 1, \varphi_0 = \pi/2, \varphi_1 = \pi/6$, unit m,kg,rad.

We also compare the accuracy using different methods.

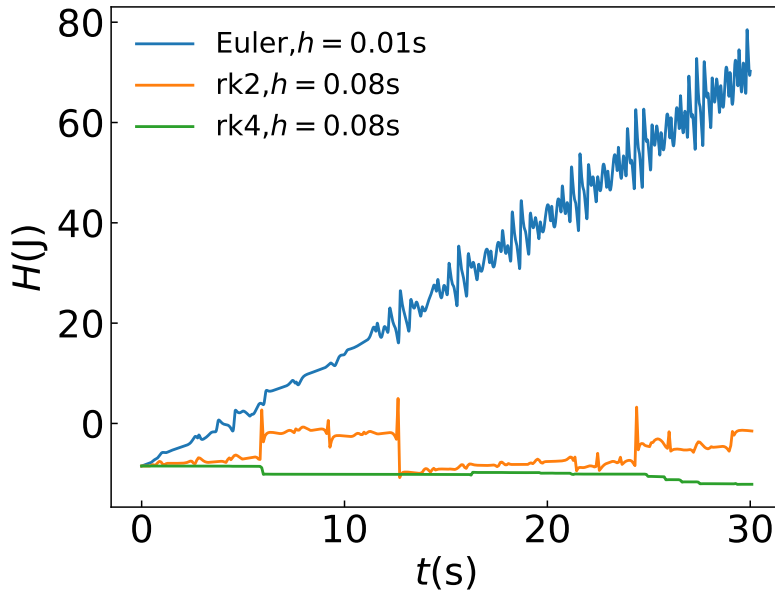


Figure 3: Comparison of accuracy among different methods. Higher order methods have better accuracy even with larger steps. Initial condition is $R_0 = 2, R_1 = 1, m_0 = 2, m_1 = 1, \varphi_0 = \pi/2, \varphi_1 = \pi/6$, unit m,kg,rad.

We plot animations for references. They are attached to the submission.

5 Multiple Pendulum

We write down the Lagrangian and Euler-Lagrange formula for multiple pendulum with n points.

$$\begin{aligned}
K &= \sum_{i=1}^n \left(\frac{1}{2} m_i \left(\left(\sum_{j=1}^i R_j \dot{\varphi}_j \cos \varphi_j \right)^2 + \left(\sum_{j=1}^i R_j \dot{\varphi}_j \sin \varphi_j \right)^2 \right) \right) \\
V &= \sum_{i=1}^n \sum_{j=1}^i -g m_i R_j \cos \varphi_j \\
\mathcal{L} &= K - V \\
\frac{\partial \mathcal{L}}{\partial \dot{\varphi}_k} &= \frac{\partial K}{\partial \dot{\varphi}_k} = \sum_{i=k}^n m_i \left(R_k \cos \varphi_k \sum_{j=1}^i R_j \dot{\varphi}_j \cos \varphi_j + R_k \sin \varphi_k \sum_{j=1}^i R_j \dot{\varphi}_j \sin \varphi_j \right) \\
&= \sum_{i=k}^n m_i \sum_{j=1}^i R_j R_k \dot{\varphi}_j \cos (\varphi_j - \varphi_k) \\
\frac{\partial \mathcal{L}}{\partial \varphi_k} &= \sum_{i=k}^n m_i \left(-R_k \dot{\varphi}_k \sin \varphi_k \sum_{j=1}^i R_j \dot{\varphi}_j \cos \varphi_j + R_k \dot{\varphi}_k \cos \varphi_k \sum_{j=1}^i R_j \dot{\varphi}_j \sin \varphi_j \right) \\
&\quad - \sum_{i=k}^n g m_i R_k \sin \varphi_k \\
&= \sum_{i=k}^n m_i \sum_{j=1}^i R_j R_k \dot{\varphi}_j \dot{\varphi}_k \sin (\varphi_j - \varphi_k) - \sum_{i=k}^n g m_i R_k \sin \varphi_k \\
\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\varphi}_k} &= \sum_{i=k}^n m_i \left(\sum_{j=1}^i R_j R_k \ddot{\varphi}_j \cos (\varphi_j - \varphi_k) - \sum_{j=1}^i R_j R_k \dot{\varphi}_j (\dot{\varphi}_j - \dot{\varphi}_k) \sin (\varphi_j - \varphi_k) \right) \\
\frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{\varphi}_k} - \frac{\partial \mathcal{L}}{\partial \varphi_k} &= 0 \\
\sum_{i=k}^n m_i \left(\sum_{j=1}^i R_j R_k \ddot{\varphi}_j \cos (\varphi_j - \varphi_k) - \sum_{j=1}^i R_j R_k \dot{\varphi}_j^2 \sin (\varphi_j - \varphi_k) + g R_k \sin \varphi_k \right) &= 0 \\
\sum_{i=k}^n m_i \left(\sum_{j=1}^i R_j \ddot{\varphi}_j \cos (\varphi_j - \varphi_k) - \sum_{j=1}^i R_j \dot{\varphi}_j^2 \sin (\varphi_j - \varphi_k) + g \sin \varphi_k \right) &= 0
\end{aligned}$$

Since φ_i and $\dot{\varphi}_i$ are known, it is a linear equation to $\ddot{\varphi}_i$, i.e. $a_k^j \ddot{\varphi}_j = b_k$.

We modify our codes as follows,

```

1 # xi = [phi_0, phi_1, ..., phi_n; phi_dot_0, phi_dot_1, ..., phi_dot_n]
2 g = 9.8
3
4 class multi_pendulum:

```

```

5  def __init__(self, N, R, M, Phi0, ode = leapfrog, step = 0.01) -> None:
6      self.N = N
7      self.R = R
8      self.M = M
9      self.Phi0 = Phi0
10     self.xi0 = np.zeros(2*N)
11     self.xi0[:N] = np.array(Phi0)
12     self.ode = ode
13     self.step = step
14
15     def diff_xi(self, xi, t):
16         N = self.N
17         # xi = [phi0, ..., phi_n; phi_dot_0, ..., phi_dot_n]
18         # diff_xi = [phi_dot_0, ..., phi_dot_n; phi_ddot_0, ..., phi_ddot_n]
19         diff_xi = np.zeros(2*N)
20         diff_xi[:N] = xi[N:]
21         phi = xi[:N]
22         phi_dot = xi[N:]
23         m = self.M
24         R = self.R
25
26         a = np.zeros([N, N])
27         b = np.zeros(N)
28         # summation here follows the equation we already derived
29         # the subscripts we use are the same as those in the derivation
30         for k in range(N):
31             for i in range(k, N):
32                 b[k] += -g * m[i] * np.sin(phi[k])
33                 for j in range(i + 1):
34                     a[k, j] += m[i] * R[j] * np.cos(phi[j] - phi[k])
35                     b[k] += m[i] * R[j] * (phi_dot[j]**2) * np.sin(phi[j] - phi[k])
36         # solve linear equation for Ax = b, A = a[k, j], b = b[k]
37         diff_xi[N:] = np.linalg.solve(a, b)
38
39         return diff_xi
40
41     def Ek(self, xi):
42         N = self.N
43         phi = xi[:N]
44         phi_dot = xi[N:]
45         m = self.M
46         R = self.R
47
48         K = 0
49         vx = vy = 0
50         for i in range(N):
51             vx += R[i] * phi_dot[i] * np.cos(phi[i])
52             vy += R[i] * phi_dot[i] * np.sin(phi[i])

```

```

53         K += 0.5 * m[i] * (vx**2 + vy**2)
54
55     return K
56
57 def V(self, xi):
58     N = self.N
59     phi = xi[:N]
60     m = self.M
61     R = self.R
62
63     V = 0
64     m_acc = np. sum(m)
65     for i in range(N):
66         V += -g * m_acc * R[i] * np.cos(phi[i])
67         m_acc -= m[i]
68
69     return V
70
71 def Lagrangian(self, xi):
72     return self.Ek(xi) - self.V(xi)
73
74 def Hamiltonian(self, xi):
75     return self.Ek(xi) + self.V(xi)
76
77 def simulate(self, t0, t1, h = 0.1):
78     T, Xi = self.ode(self.diff_xi, t0, self.xi0, t1, h)
79     return T, Xi

```

Again, we verify the convergence of total energy.

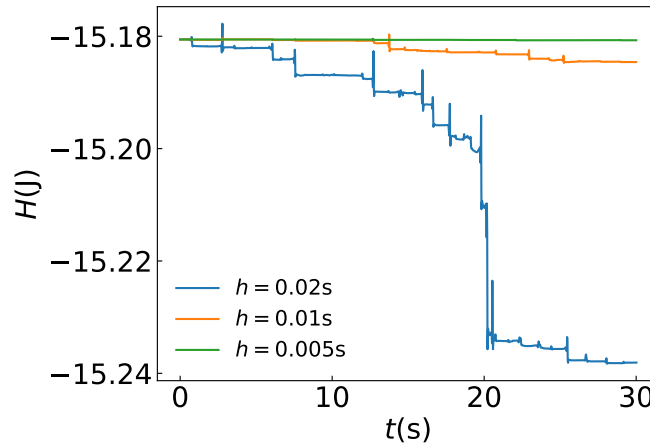


Figure 4: Hamiltonian versus time using rk4 with decreasing steps. With smaller step, the total energy converges. Initial condition is $N = 3, R_0 = 2, R_1 = 1, R_2 = 1, m_0 = 2, m_1 = 1, m_2 = 0.5, \varphi_0 = \pi/2, \varphi_1 = \pi/6, \varphi_2 = -\pi/3$, unit m,kg,rad.

Here we plot some trajectories of the pendulum.

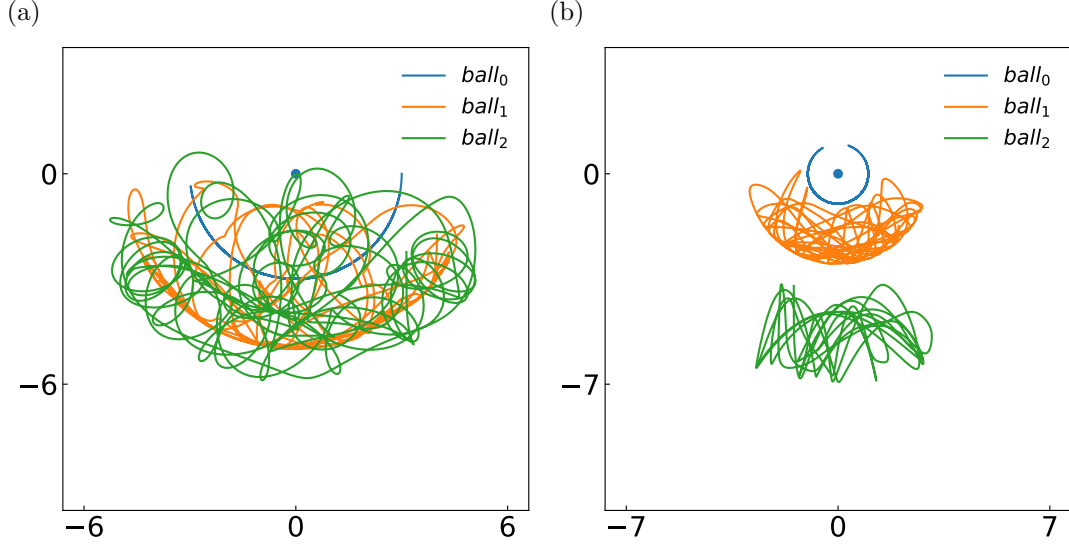


Figure 5: $N = 3$. (a) Initial condition $R_0 = 3, R_1 = 2, R_2 = 1, m_0 = 2, m_1 = 1, m_2 = 0.5, \varphi_0 = \pi/2, \varphi_1 = \pi/6, \varphi_2 = -\pi/3$, unit m,kg,rad. (b) Initial condition $R_0 = 1, R_1 = 2, R_2 = 4, m_0 = 2, m_1 = 1, m_2 = 3, \varphi_0 = \pi/2, \varphi_1 = \pi/6, \varphi_2 = -\pi/3$, unit m,kg,rad.

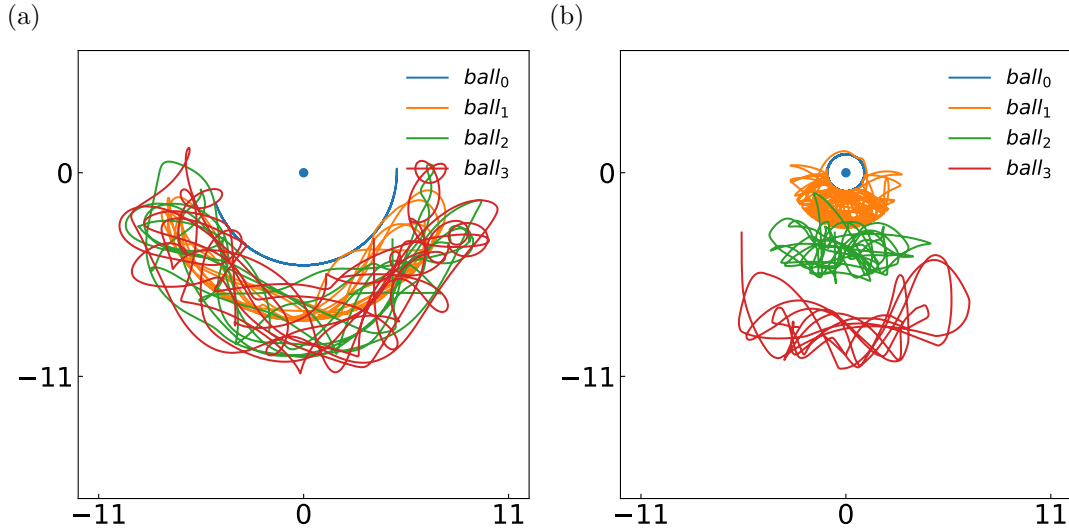


Figure 6: $N = 4$. (a) Initial condition $R_0 = 5, R_1 = 3, R_2 = 2, R_3 = 1, m_0 = 2, m_1 = 1, m_2 = 0.5, m_3 = 0.25, \varphi_0 = \pi/2, \varphi_1 = \pi/6, \varphi_2 = -\pi/3, \varphi_3 = -\pi/2$, unit m,kg,rad. (b) Initial condition $R_0 = 1, R_1 = 2, R_2 = 3, R_3 = 5, m_0 = 1, m_1 = 2, m_2 = 4, m_3 = 8, \varphi_0 = \pi/2, \varphi_1 = \pi/6, \varphi_2 = -\pi/3, \varphi_3 = -\pi/2$, unit m,kg,rad.