

### Problem 1

$$SO(n) = \{ A \in \mathbb{R}^{n \times n} \mid A^T A = I_{n \times n} \text{ and } \det(A) = 1 \}$$

$$A = \frac{d}{dt}(R)R^{-1} = \dot{R}R^T \quad A^T = R\dot{R}^T$$

↳ skew symmetric means that  $A^T = -A$  and if

$R \in SO(n)$ , then  $R^T R = I_{n \times n}$  and  $\det(R) = 1$

Since we know  $R^T R = R R^T = I_{n \times n}$ ,  $\frac{d}{dt}(R R^T) = 0$

$$\frac{d}{dt}(R R^T) = 0$$

$$\dot{R}R^T + R\dot{R}^T = 0$$

$$\dot{R}R^T = -R\dot{R}^T$$

where  $A = \dot{R}R^T$  and  $A^T = (\dot{R}R^T)^T = R\dot{R}^T$

such that  $\Rightarrow \dot{R}R^T = -R\dot{R}^T$

$$\Rightarrow A = -A^T$$

### Problem 2

$$\hat{w}r_b = -\hat{r}_b w$$

$$w \times r = -(r \times w)$$

$$\begin{vmatrix} w_1 \\ w_2 \\ w_3 \end{vmatrix} \times \begin{vmatrix} r_1 \\ r_2 \\ r_3 \end{vmatrix} = - \begin{vmatrix} r_1 \\ r_2 \\ r_3 \end{vmatrix} \times \begin{vmatrix} w_1 \\ w_2 \\ w_3 \end{vmatrix}$$

$$\begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ w_1 & w_2 & w_3 \\ r_1 & r_2 & r_3 \end{vmatrix} = - \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ r_1 & r_2 & r_3 \\ w_1 & w_2 & w_3 \end{vmatrix}$$

$$\Rightarrow (w_2 r_3 - w_3 r_2)\hat{i} - (w_1 r_3 - w_3 r_1)\hat{j} + (w_1 r_2 - w_2 r_1)\hat{k} - [(r_2 w_3 - r_3 w_2)\hat{i} - (r_1 w_3 - r_3 w_1)\hat{j} + (r_1 w_2 - r_2 w_1)\hat{k}]$$

$$\Rightarrow \begin{vmatrix} \omega_2 r_3 - \omega_3 r_2 \\ \omega_3 r_1 - \omega_1 r_3 \\ \omega_1 r_2 - \omega_2 r_1 \end{vmatrix} = - \begin{vmatrix} r_2 \omega_3 - r_3 \omega_2 \\ r_3 \omega_1 - r_1 \omega_3 \\ r_1 \omega_2 - r_2 \omega_1 \end{vmatrix}$$

$$\Rightarrow \begin{vmatrix} \omega_2 r_3 - \omega_3 r_2 \\ \omega_3 r_1 - \omega_1 r_3 \\ \omega_1 r_2 - \omega_2 r_1 \end{vmatrix} = \begin{vmatrix} r_3 \omega_2 - r_2 \omega_3 \\ r_1 \omega_3 - r_3 \omega_1 \\ r_2 \omega_1 - r_1 \omega_2 \end{vmatrix}$$

$$\Rightarrow \hat{\omega} r_b = -\hat{r}_b \omega$$



of the two legs,  $\theta_1$  and  $\theta_2$  are the angles between the legs and the green vertical dash line. The feet are constrained on the ground, and there is no friction between the feet and the ground.

Each leg is a rigid body with length  $L = 1$ , width  $W = 0.2$ , mass  $m = 1$ , and rotational inertia  $J = 1$  (assuming the center of mass is at the center of geometry). Moreover, there are two torques applied on  $\theta_1$  and  $\theta_2$  to control the legs to track a desired trajectory:

$$\begin{aligned}\theta_1^d(t) &= \frac{\pi}{15} + \frac{\pi}{3} \sin^2\left(\frac{t}{2}\right) \\ \theta_2^d(t) &= -\frac{\pi}{15} - \frac{\pi}{3} \sin^2\left(\frac{t}{2}\right)\end{aligned}$$

and the torques are:

$$\begin{aligned}F_{\theta_1} &= -k_1(\theta_1 - \theta_1^d) \\ F_{\theta_2} &= -k_1(\theta_2 - \theta_2^d)\end{aligned}$$

In this problem we use  $k_1 = 20$ .

Given the model description above, define the frames that you need (several example frames are shown in the image as well), simulate the motion of the biped from rest for  $t \in [0, 10]$ ,  $dt = 0.01$ , with initial condition  $q = [0, L_1 \cos(\frac{\pi}{15}), \frac{\pi}{15}, -\frac{\pi}{15}]$ . You will need to modify the animation function to display the leg as a rectangle, an example of the animation can be found at <https://youtu.be/w8XHYrEoWTc>.

*Hint 1: Even though this is a 2D system, in order to compute kinetic energy from both translation and rotation you will need to model the system in the 3D world --- the z coordinate is always zero and the rotation is around the z axis (based on these facts, what should the SE(3) matrix and rotational inertia tensor look like?). This also means you need to represent transformations in SE(3) and the body velocity  $\mathcal{V}_b \in \mathbb{R}^6$ .*

*Hint 2: It could be helpful to define several helper functions for all the matrix operations you will need to use. For example, a function that returns SE(3) matrices given a rotation angle and 2D translation vector, functions for "hat" and "unhat" operations, a function for the matrix inverse of SE(3) (which is definitely not the same as the SymPy matrix inverse function), and a function that returns the time derivative of SO(3) or SE(3).*

*Hint 3: In this problem the external force depends on time  $t$ . Therefore, in order to solve for the symbolic solution you need to substitute your configuration variables, which are defined as symbolic functions of time  $t$  (such as  $\theta_1(t)$  and  $\frac{d}{dt}\theta_1(t)$ ), with dummy symbolic variables. For the same reason (the dynamics now explicitly depend on time), you will need to do some tiny modifications on the "integrate" and "simulate" functions, a good reference can be found at [https://en.wikipedia.org/wiki/Runge-Kutta\\_methods](https://en.wikipedia.org/wiki/Runge-Kutta_methods).*

*Hint 4: Symbolically solving this system should be fast, but if you encountered some problem when solving the dynamics symbolically, an alternative method is to solve the system numerically --- substitute in the system state at each time step during simulation and solve for the numerical solution --- but based on my experience, this would cost more than one hour for 500 time steps, so it's not recommended.*

*Hint 5: The animation of this problem is similar to the one in last homework --- the coordinates of the vertices in the body frame are constant, you just need to transfer them back to the world frame using the transformation matrices you already have in the simulation.*

*Hint 6: Be careful to consider the relationship between the frames and to not build in any implicit assumptions (such as assuming some variables are fixed).*

*Hint 7: The rotation, by convention, is assumed to follow the right hand rule, which means the z-axis is out of the screen and the positive rotation orientation is counter-clockwise. Make sure you follow a consistent set of positive directions for all the computation.*

*Hint 8: This problem is designed as a "mini-project", it could help you estimate the complexity of your final project, and you could adjust your proposal based on your experience with this problem.*

**Turn in: A copy of the code used to simulate and animate the system. Also, include a plot of the trajectory and upload a video of the animation separately through Canvas. The video should be in ".mp4" format, you can use screen capture or record the screen directly with your phone.**

```
1 from os import get_blocking
2 from sympy import sin, cos
3 import sympy as sym
4 import numpy as np
5 import matplotlib.pyplot as plt
6 from IPython.display import Markdown, display
7
8 # Defining necessary symbols
9 t = sym.symbols('t')
10 l = sym.symbols('l')
11 m = sym.symbols('m')
12 g = sym.symbols('g')
13 J = sym.symbols('J')
14 k = sym.symbols('k')
15
16
17 # define configuration variables
18 x = sym.Function('x')(t)
19 y = sym.Function('y')(t)
```

```

20
21 theta1 = sym.Function('theta_1')(t)
22 theta2 = sym.Function('theta_2')(t)
23
24 # derivatives of configuration variables
25 x_d = x.diff(t)
26 y_d = y.diff(t)
27 theta1_d = theta1.diff(t)
28 theta2_d = theta2.diff(t)
29
30 x_dd = x_d.diff(t)
31 y_dd = y_d.diff(t)
32 theta1_dd = theta1_d.diff(t)
33 theta2_dd = theta2_d.diff(t)
34
35 ### ----- DEFINING TRANSFORMATIONS ----- ###
36
37
38 # Defining Function to compute SE(3) given rotation angle and translation.
39 def SE3(theta, p):
40     assert len(p) == 3
41     if theta == 0:
42         g = sym.BlockMatrix([[sym.eye(3), p], [sym.zeros(1, 3), sym.Matrix([1])]])
43         g = g.as_explicit()
44     else:
45         R = sym.Matrix([ [cos(theta), -sin(theta), 0], [sin(theta), cos(theta), 0], [0, 0, 1] ])
46         Rot = sym.BlockMatrix([[R, sym.zeros(3, 1)], [sym.zeros(1, 3), sym.Matrix([1])]])
47         p = sym.BlockMatrix([[sym.eye(3), p], [sym.zeros(1, 3), sym.Matrix([1])]])
48         g = Rot*p
49         g = g.as_explicit()
50     return g
51
52 ### DEFINING RIGID BODY TRANSFORMATIONS
53 g_wa = SE3(0, sym.Matrix([x, y, 0]))
54 g_ab = SE3(theta1, sym.Matrix([0, -1/2, 0]))
55 g_ac = SE3(theta2, sym.Matrix([0, -1/2, 0]))
56 g_bd = SE3(0, sym.Matrix([0, -1/2, 0]))
57 g_ce = SE3(0, sym.Matrix([0, -1/2, 0]))
58
59 # Defining function to compute the derivative of an SE3
60 def derivate(m):
61     return m.diff(t)
62
63 # Defining a function to compute the inverse of an SE3
64 def inverse(m):
65     temp = sym.zeros(4, 4)
66     temp[0:3, 0:3] = m[0:3, 0:3].T
67     temp[0:3, 3] = -temp[0:3, 0:3] * m[0:3, 3]
68     temp[3, 0:3] = sym.zeros(1, 3)
69     temp[3, 3] = 1
70     return temp
71
72 # Defining a function that convert a hatted 4x4 matrix into an unhatted 6x1 vector
73 def unhat(m):
74     assert m.shape == (4, 4)
75     w_hat = m[0:3, 0:3]
76     w = sym.Matrix([w_hat[2, 1], -w_hat[2, 0], w_hat[1, 0]])
77     v = m[0:3, 3] # Select the first three elements of the fourth column
78     unhat = sym.Matrix.vstack(v, w)
79     return unhat
80
81 # Defining a function that convert an unhatted 6x1 vector into a hatted 4x4 matrix
82 def hat(m):
83     assert m.shape == (6, 1)
84     v = m[0:3, 0]
85     w = m[3:6, 0]
86     w_hat = sym.Matrix([[0, -w[2], w[1]], [w[2], 0, -w[0]], [-w[1], w[0], 0]])
87     hat = sym.BlockMatrix([[w_hat, v], [sym.zeros(1, 3), sym.zeros(1, 1)]])
88     hat = hat.as_explicit()
89     return hat
90
91 # DEFINING BODY VELOCITIES
92 # define the transformation from W --> C
93 g_wc = g_wa*g_ac
94 # define body velocity at c
95 v_c_hat = inverse(g_wc)*derivate(g_wc)
96 # unhat body velocity

```

```

97 Vc = unhat(Vc_hat)
98
99 # define the transformation from W --> B
100 g_wb = g_wa*g_ab
101 # define the body velocity at b
102 Vb_hat = inverse(g_wb)*derivate(g_wb)
103 # unhat the body velocity
104 Vb = unhat(Vb_hat)
105
106
107 # Defining the rotational Inertia Tensors
108 # there is only rotational inertia in the z-direction
109 I_rot_c = sym.Matrix([[0, 0, 0], [0, 0, 0], [0, 0, J]])
110 # for the rectangle with COM at b, it has the same rotational inertia
111 I_rot_b = sym.Matrix([[0, 0, 0], [0, 0, 0], [0, 0, J]])
112
113 # defining the dynamic matrix
114 dyn_mat = sym.BlockMatrix( [ [m*sym.eye(3), sym.zeros(3,3)], [sym.zeros(3,3), I_rot_c] ] )
115 dyn_mat = dyn_mat.as_explicit()
116
117 # DEFINING THE KINETIC ENERGY #
118 KE = 0.5*(Vb.T)*dyn_mat*Vb + 0.5*(Vc.T)*dyn_mat*Vc
119
120 # DEFINING THE POTENTIAL ENERGY #
121 rc_bar = sym.Matrix([0, 0, 0, 1])
122 rwc_bar = g_wc*rc_bar
123 hc = sym.Matrix([0, 1, 0, 0]).T*rwc_bar
124
125 rb_bar = sym.Matrix([0, 0, 0, 1])
126 rwb_bar = g_wb*rb_bar
127 hb = sym.Matrix([0, 1, 0, 0]).T*rwb_bar
128
129 PE = m*g*hc + m*g*hb
130
131
132 # LAGRANGIAN
133 L = KE - PE
134
135 # determening forcing terms and constraints
136
137 # CONSTRAINTS
138 # defining the transformation from the world frame to frame d
139 g_wd = g_wa*g_ab*g_bd
140 # position of the center of frame d wrt frame d
141 rd_bar = sym.Matrix([0, 0, 0, 1])
142 # defining the coordinate change from w to d
143 rwd_bar = g_wd*rd_bar
144 # height of center of frame d wrt world frame
145 hd = sym.Matrix([0, 1, 0, 0]).T*rwd_bar
146 phi1 = hd[0] # the height of the center of d wrt to the world frame has to be 0.
147
148 # defining the transformation from the world frame to frame e
149 g_we = g_wa*g_ac*g_ce
150 # position of the center of frame e wrt frame e
151 re_bar = sym.Matrix([0, 0, 0, 1])
152 # defining the coordinate change from w to e
153 rwe_bar = g_we*re_bar
154 # height of center of frame e wrt world frame
155 he = sym.Matrix([0, 1, 0, 0]).T*rwe_bar
156 phi2 = he[0] # the height of the center of e wrt to the world frame has to be 0.
157
158 lamb1 = sym.symbols('lambda_1')
159 lamb2 = sym.symbols('lambda_2')
160
161 # determening the gradients of constraint 1 for each configuration
162 dphi1_dx = phi1.diff(x)
163 dphi1_dy = phi1.diff(y)
164 dphi1_dtheta1 = phi1.diff(theta1)
165 dphi1_dtheta2 = phi1.diff(theta2)
166
167 # determening the gradients of constraint 2 for each configuration
168 dphi2_dx = phi2.diff(x)
169 dphi2_dy = phi2.diff(y)
170 dphi2_dtheta1 = phi2.diff(theta1)
171 dphi2_dtheta2 = phi2.diff(theta2)
172
173

```

```

174 # FORCING
175 # desired trajectories
176 thetal_desired = np.pi/15 + (np.pi/3)*(sin(t/2))**2
177 theta2_desired = -np.pi/15 - (np.pi/3)*(sin(t/2))**2
178
179 # torques
180 F1 = -k*(thetal - thetal_desired)
181 F2 = -k*(theta2 - theta2_desired)
182
183
184 ## EULER-LAGRANGE and CONSTRAINT EQUATIONS
185
186 # define subs
187 subs = {m: 1, k:20, J:1, l:1, g:9.8}
188
189 # lhs of EL and constraint
190 lhs = sym.Matrix([ (L.diff(x_d).diff(t) - L.diff(x)).simplify(), (L.diff(y_d).diff(t) - L.diff(y)).simplify(), (L.diff(thetal_c
191
192 # rhs of EL and constraint
193 # only forcing on thetal and theta2
194 rhs = sym.Matrix([ (lamb1*dphil_dx + lamb2*dphi2_dx), (lamb1*dphil_dy + lamb2*dphi2_dy), (lamb1*dphil_dthetal + lamb2*dphi2_dt
195 # combine into equation to get euler lagrange
196 eqs = sym.Eq(lhs.subs(subs), rhs.subs(subs))
197 # determine what we are solving for
198 c = [lamb1, lamb2, x_dd, y_dd, thetal_dd, theta2_dd]
199
200 soln = sym.solve(eqs, c)
201 display(soln)
202

```

$$\lambda_1 : -\frac{8.0 \cdot 10^{15} \theta_1(t) \sin(\theta_1(t)) \sin^2(\theta_2(t))}{10000000000000.0 \sin^2(\theta_1(t)) \cos^2(\theta_2(t)) - 1.0 \cdot 10^{15} \sin^2(\theta_1(t)) + 20000000000000.0 \sin(\theta_1(t)) \sin(\theta_2(t)) \cos(\theta_1(t)) \cos(\theta_2(t)) + 1000}$$

```

1 xddot = soln[x_dd]
2 yddot = soln[y_dd]
3 thetalddot = soln[thetal_dd]
4 theta2ddot = soln[theta2_dd]
5
6 xddot_simplified = sym.simplify(xddot)
7 yddot_simplified = sym.simplify(yddot)
8 thetalddot_simplified = sym.simplify(thetalddot)
9 theta2ddot_simplified = sym.simplify(theta2ddot)
10
11 l1 = sym.lambdify([x, y, thetal, theta2, x_d, y_d, thetal_d, theta2_d, t], xddot_simplified, "sympy")
12 l2 = sym.lambdify([x, y, thetal, theta2, x_d, y_d, thetal_d, theta2_d, t], yddot_simplified, "sympy")
13 l3 = sym.lambdify([x, y, thetal, theta2, x_d, y_d, thetal_d, theta2_d, t], thetalddot_simplified, "sympy")
14 l4 = sym.lambdify([x, y, thetal, theta2, x_d, y_d, thetal_d, theta2_d, t], theta2ddot_simplified, "sympy")
15
16
17 ### PLOTTING TRAJECTORY ###
18 def integrate(f, xt, dt, t):
19     """
20     This function takes in an initial condition x(t) and a timestep dt,
21     as well as a dynamical system f(x) that outputs a vector of the
22     same dimension as x(t). It outputs a vector x(t+dt) at the future
23     time step.
24
25     Parameters
26     =====
27     dyn: Python function
28         derivate of the system at a given step x(t),
29         it can considered as \dot{x}(t) = func(x(t))
30     xt: NumPy array
31         current step x(t)
32     dt:
33         step size for integration
34
35     Return
36     =====
37     new_xt:
38         value of x(t+dt) integrated from x(t)
39     """
40     k1 = dt * f(xt, t)
41     k2 = dt * f(xt+k1/2., t)
42     k3 = dt * f(xt+k2/2., t)

```

```

43     k4 = dt * f(xt+k3, t)
44     new_xt = xt + (1/6.) * (k1+2.0*k2+2.0*k3+k4)
45     return new_xt
46
47 def simulate(f, x0, tspan, dt, integrate):
48     """
49     This function takes in an initial condition x0, a timestep dt,
50     a time span tspan consisting of a list [min_time, max_time],
51     as well as a dynamical system f(x) that outputs a vector of the
52     same dimension as x0. It outputs a full trajectory simulated
53     over the time span of dimensions (xvec_size, time_vec_size).
54
55     Parameters
56     =====
57     f: Python function
58         derivate of the system at a given step x(t),
59         it can considered as  $\dot{x}(t) = \text{func}(x(t))$ 
60     x0: NumPy array
61         initial conditions
62     tspan: Python list
63         tspan = [min_time, max_time], it defines the start and end
64         time of simulation
65     dt:
66         time step for numerical integration
67     integrate: Python function
68         numerical integration method used in this simulation
69
70     Return
71     =====
72     x_traj:
73         simulated trajectory of x(t), theta(t), xdot(t), thetadot(t) from t=0 to tf
74     """
75     N = int((max(tspan)-0)/dt)
76     x = np.copy(x0)
77     tvec = np.linspace(0,max(tspan),N)
78     xtraj = np.zeros((len(x0),N))
79     t = tspan[0]
80     for i in range(N):
81         xtraj[:,i]=integrate(f,x,dt,t)
82         x = np.copy(xtraj[:,i])
83         t = t + dt
84     return xtraj
85
86 #####
87 def xdd(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t):
88     return 0
89
90 def ydd(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t):
91     return l2(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t)
92
93 def theta1dd(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t):
94     return l3(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t)
95
96 def theta2dd(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t):
97     return l4(x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d, t)
98
99 def dyn(s, t):
100     """
101     System dynamics function (extended)
102
103     Parameters
104     =====
105     s: NumPy array
106         s = [x, y, theta1, theta2, x_d, y_d, theta1_d, theta2_d] is the extended system
107         state vector, including the position and
108         the velocity of the particle
109
110     Return
111     =====
112     sdot: NumPy array
113         time derivative of input state vector,
114         sdot = [x_d, y_d, theta1_d, theta2_d, x_dd, y_dd, theta1_dd, theta2_dd]
115     """
116     return np.array([s[4], s[5], s[6], s[7],
117                     xdd(s[0], s[1], s[2], s[3], s[4], s[5], s[6], s[7], t),
118                     ydd(s[0], s[1], s[2], s[3], s[4], s[5], s[6], s[7], t),
119                     theta1dd(s[0], s[1], s[2], s[3], s[4], s[5], s[6], s[7], t),
120                     theta2dd(s[0], s[1], s[2], s[3], s[4], s[5], s[6], s[7], t)])

```



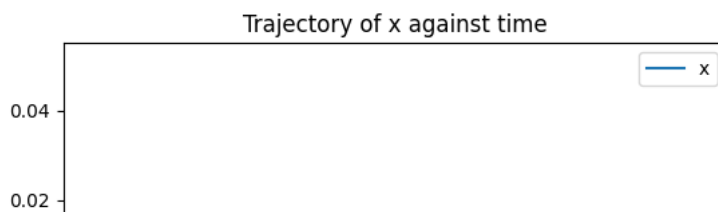
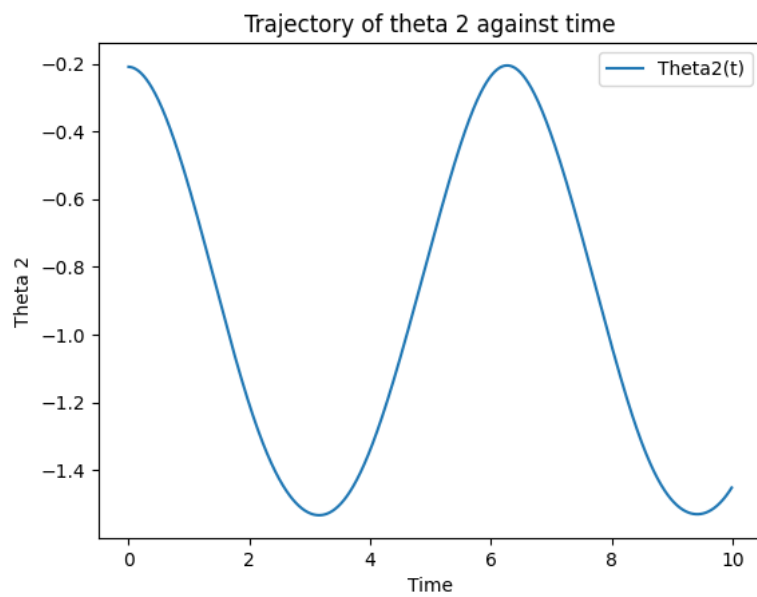
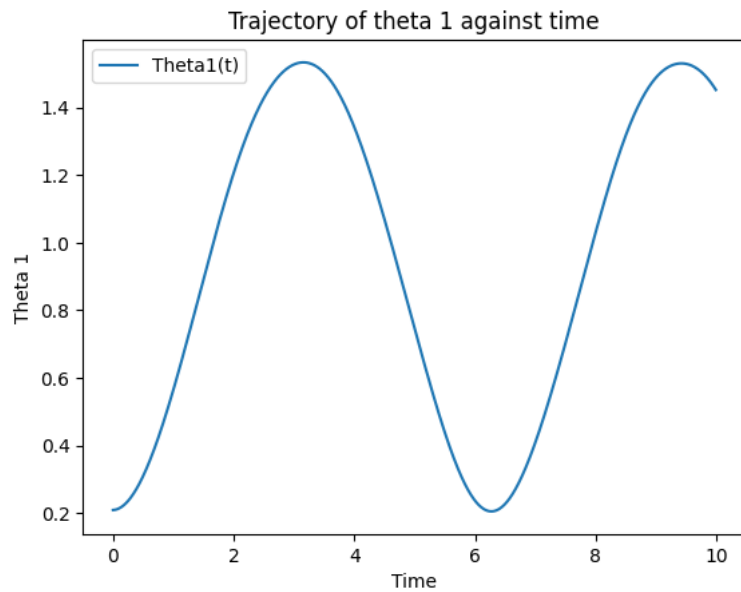
```

120         theta2_uu(s[u], s[l], s[z], s[j], s[*], s[u], s[l], u))
121
122
123 # define initial state (theta1 = theta2 = -pi/2 , theta1_d = theta2_d = 0)
124 # define initial state
125 x0 = 0
126 y0 = 1*np.cos(np.pi/15)
127 theta1_0 = np.pi/15
128 theta2_0 = -np.pi/15
129 x_d_0 = 0
130 y_d_0 = 0
131 theta1_d_0 = 0
132 theta2_d_0 = 0
133 s0 = np.array([x0, y0, theta1_0, theta2_0, x_d_0, y_d_0, theta1_d_0, theta2_d_0])
134 # simulat from t=0 to 10, since dt=0.01, the returned trajectory
135 # will have 10/0.01=1000 time steps, each time step contains extended
136 # system state vector [x(t), xdot(t)]
137 traj = simulate(dyn, s0, [0, 10], 0.01, integrate)
138
139 import matplotlib.pyplot as plt
140
141 # determening trajectories of configuration variables and their derivatives
142
143 #traj = simulate(dyn, s0, [0, 10], 0.01, integrate)
144
145 x_traj = traj[0]
146 y_traj = traj[1]
147 theta1_traj = traj[2]
148 theta2_traj = traj[3]
149 x_d_traj = traj[4]
150 y_d_traj = traj[5]
151 theta1_d_traj = traj[6]
152 theta2_d_traj = traj[7]
153
154
155 # initial state is already defined above and is part of theta 1 and theta 2 trajectories
156
157 # time step array
158 # 3/0.01 will give 300 time steps
159 timespan = np.arange(0, 10, 0.01)
160
161
162 plt.plot(timespan, theta1_traj, label='Theta1(t)')
163 plt.title("Trajectory of theta 1 against time")
164 plt.ylabel('Theta 1')
165 plt.xlabel('Time')
166 plt.legend()
167 plt.show()
168
169 plt.plot(timespan, theta2_traj, label='Theta2(t)')
170 plt.title("Trajectory of theta 2 against time")
171 plt.ylabel('Theta 2')
172 plt.legend()
173 plt.xlabel('Time')
174 plt.show()
175
176 plt.plot(timespan, x_traj, label='x')
177 plt.title("Trajectory of x against time")
178 plt.ylabel('x_traj')
179 plt.xlabel('Time')
180 plt.legend()
181 plt.show()
182
183 plt.plot(timespan, y_traj, label='y')
184 plt.title("Trajectory of y against time")
185 plt.ylabel('y_traj')
186 plt.legend()
187 plt.xlabel('Time')
188 plt.show()
189
190
191

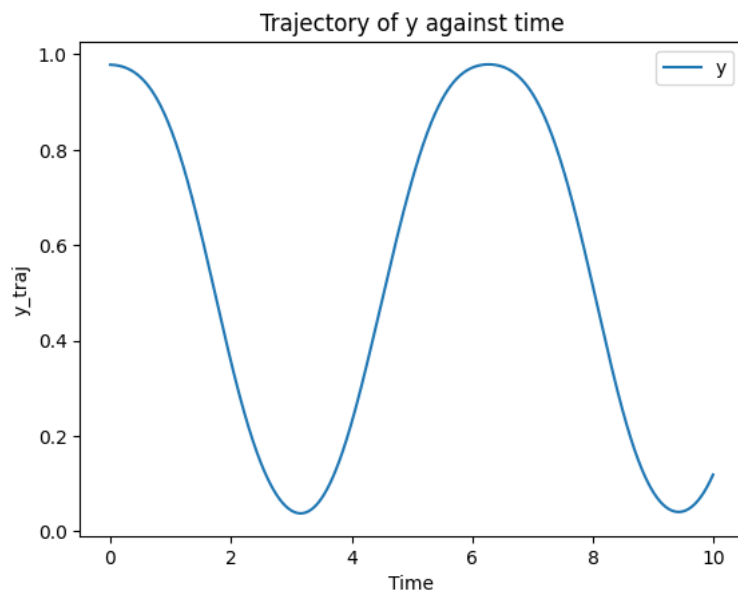
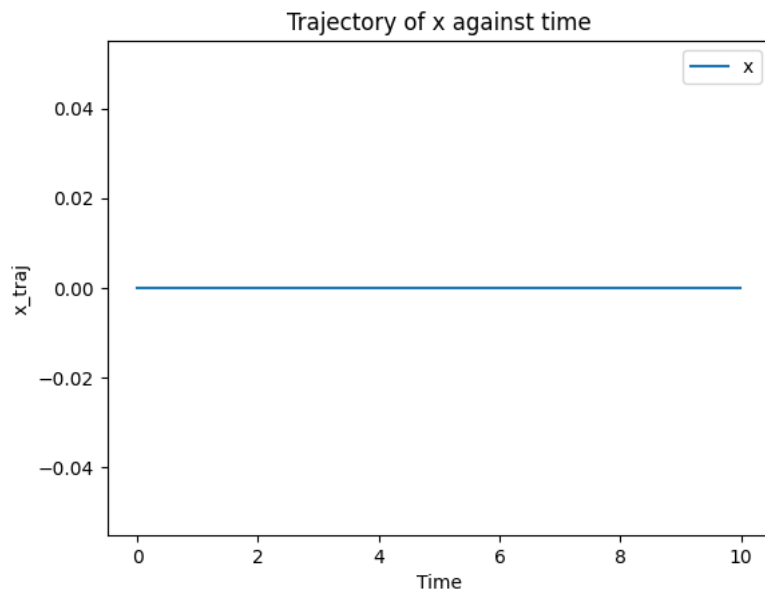
```



Shape of traj: (8, 1000)



198



```

1 def animate_split(theta_array,L1=1,L2=1,T=10):
2     """
3     Function to generate web-based animation of a person doing a split
4
5     Parameters:
6     =====
7     theta_array:
8         trajectory of theta1 and theta2, should be a NumPy array with
9         shape of (2,N)
10    L1:
11        length of the first leg
12    L2:
13        length of the second leg
14    T:
15        length/seconds of animation duration
16
17    Returns: None
18    """
19
20    #####
21    # Imports required for animation.
22    from plotly.offline import init_notebook_mode, iplot
23    from IPython.display import display, HTML
24    import plotly.graph_objects as go
25

```

```

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23    from IPython.display import display, HTML
24    import plotly.graph_objects as go
25
26    #####
27    # Browser configuration.
28    def configure_plotly_browser_state():
29        import IPython
30        display(IPython.core.display.HTML('''
31            <script src="/static/components/requirejs/require.js"></script>
32            <script>
33                requirejs.config({
34                    paths: {
35                        base: '/static/base',
36                        plotly: 'https://cdn.plot.ly/plotly-1.5.1.min.js?noext',
37                    },
38                });
39            </script>
40            '''))
41    configure_plotly_browser_state()
42    init_notebook_mode(connected=False)
43
44    #####
45    # Getting data from split angle trajectories.
46    x1 = 0
47    y1 = theta_array[1]
48    xx1= L1*np.sin(theta_array[2])
49    yy1=-L1*np.cos(theta_array[2])
50    xx2=L2*np.sin(theta_array[3])
51    yy2=-L2*np.cos(theta_array[3])
52    N = len(theta_array[0]) # Need this for specifying length of simulation
53
54    #####
55    # Using these to specify axis limits.
56    xm = -3 #np.min(xx1)-0.5
57    xM = 3 #np.max(xx1)+0.5
58    ym = -3 #np.min(yy1)-2.5
59    yM = 3 #np.max(yy1)+1.5
60
61    #####
62    # Defining data dictionary

```

```

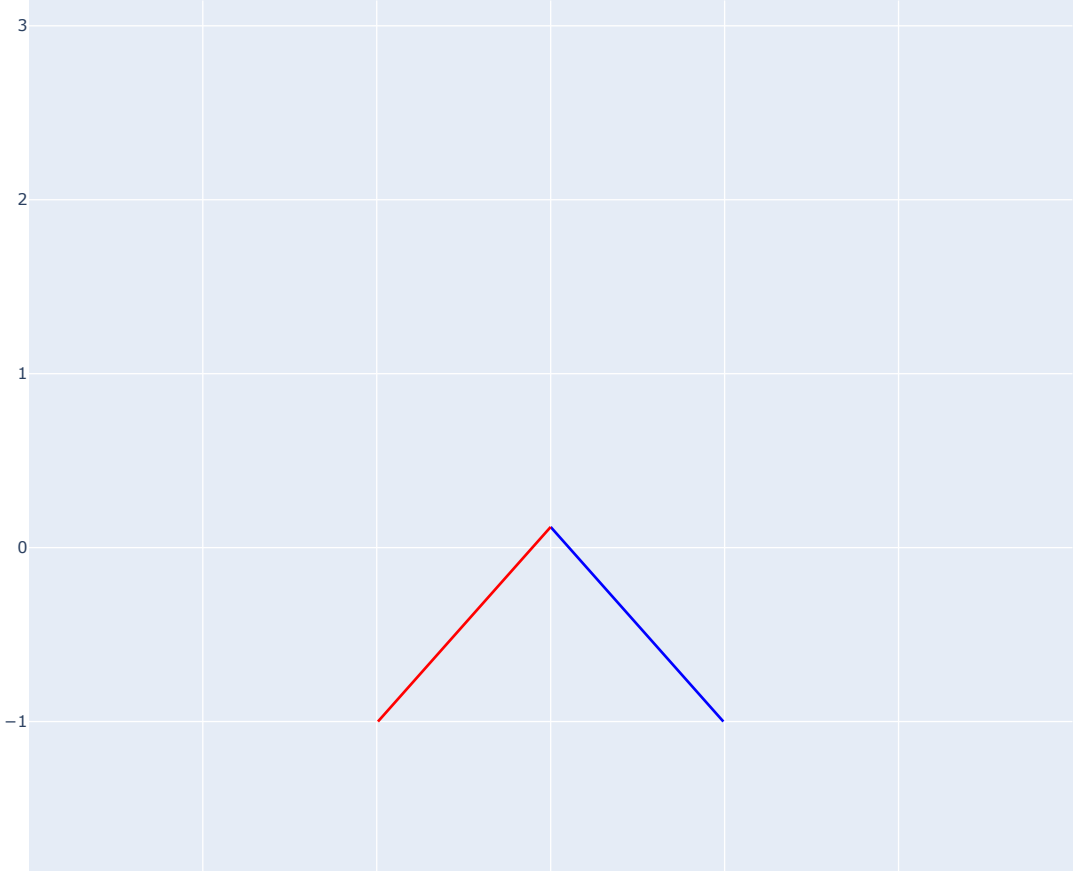
62 # Defining data dictionary.
63 # Trajectories are here.
64 data=[
65     # first two objects correspond to the legs
66     dict(name='Leg 1'),
67     dict(name='Leg 2'),
68 ]
69
70 #####
71 # Preparing simulation layout.
72 # Title and axis ranges are here.
73 layout=dict(autosize=False, width=1000, height=1000,
74             xaxis=dict(range=[xm, xM], autorange=False, zeroline=False, dtick=1),
75             yaxis=dict(range=[ym, yM], autorange=False, zeroline=False, scaleanchor = "x", dtick=1),
76             title='Split Simulation',
77             hovermode='closest',
78             updatemenus= [{ 'type': 'buttons',
79                             'buttons': [{ 'label': 'Play', 'method': 'animate',
80                                           'args': [None, { 'frame': { 'duration': T, 'redraw': False } }],
81                                           { 'args': [None], { 'frame': { 'duration': T, 'redraw': False }, 'mode': 'immediate',
82                                           'transition': { 'duration': 0 } }], 'label': 'Pause', 'method': 'animate' }
83                             ]
84                         })
85
86 #####
87 # Defining the frames of the simulation.
88 frames=[dict(data=[
89     # display legs in each animation frame
90     dict(x=[x1, xx1[k]],
91          y=[y1[k], -L1*np.cos(theta_array[0])[k]],
92          mode='lines', line=dict(width=2, color='blue')),
93     dict(x=[x1, xx2[k]],
94          y=[y1[k], -L1*np.cos(theta_array[0])[k]],
95          mode='lines', line=dict(width=2, color='red')),
96 ]) for k in range(N)]
97
98 #####
99 # Merging everything into one dictionary and producing the animation.
100 figure1=dict(data=data, layout=layout, frames=frames)
101 iplot(figure1)
102
103
104
105 animate_split(traj, 1, 1, 10)

```

Split Simulation

Play

Pause



Leg 1  
Leg 2

✓ 3s completed at 8:39 AM

● ✕

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