

# Assignment 1: Modeling (Solution)

University of California Berkeley

ME C231A, EE C220B, Experiential Advanced Control I

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# Question 1. Equilibrium Point and Linearization

## Part (a)

At equilibrium  $\dot{\theta}_1 = \dot{\theta}_2 = \ddot{\theta}_1 = \ddot{\theta}_2 = 0$ . We set  $\bar{\theta}_2 = 0$  and get

$$\begin{aligned} mgl \sin \theta_1 - a^2 k \cos \theta_1 \sin \theta_1 + T(t) &= 0 \\ a^2 k \sin \theta_1 + \alpha &= 0 \end{aligned}$$

After some rearrangement we get

$$\begin{aligned} \bar{\theta}_1 &= \arcsin \left( -\frac{\alpha}{a^2 k} \right) \\ \bar{T} &= mgl \frac{\alpha}{a^2 k} - \alpha \cos \arcsin \left( -\frac{\alpha}{a^2 k} \right) \end{aligned}$$

Using  $\cos \arcsin(x) = \sqrt{1 - x^2}$ , the equation for  $\bar{T}$  can be simplified to

$$\bar{T} = \frac{\alpha}{a^2 k} \left( mgl - \sqrt{a^4 k^2 - \alpha^2} \right)$$

## Part (b)

The new equilibrium point is  $\bar{\theta}_1 = \bar{\theta}_2 = \bar{T} = 0$ . We set the state and input variables as follows

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \theta_1 - \bar{\theta}_1 = \theta_1 \\ \dot{\theta}_1 \\ \theta_2 - \bar{\theta}_2 = \theta_2 \\ \dot{\theta}_2 \end{bmatrix}, \quad u = T(t) - \bar{T} = T(t)$$

The system of four 1st order ODE  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, u)$  can be written as

$$\mathbf{f}(\mathbf{x}, u) = \begin{bmatrix} x_2 \\ \frac{1}{ml^2} (mgl \sin x_1 + a^2 \cos x_1 (k(\sin x_3 - \sin x_1) + d(x_4 - x_2)) + u) \\ x_4 \\ \frac{1}{ml^2} (mgl \sin x_3 - a^2 \cos x_3 (k(\sin x_3 - \sin x_1) + d(x_4 - x_2)) + \alpha + \beta x_4^2) \end{bmatrix}$$

Taking the hint into account and linearizing with

$$A = \left. \frac{\partial \mathbf{f}(\mathbf{x}, u)}{\partial \mathbf{x}} \right|_{x_{ss}, u_{ss}}, \quad B = \left. \frac{\partial \mathbf{f}(\mathbf{x}, u)}{\partial u} \right|_{x_{ss}, u_{ss}}$$

leads to a state space description of the form  $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}u$  given by

$$\begin{aligned} \dot{\mathbf{x}}(t) &= \begin{bmatrix} 0 & 1 & 0 & 0 \\ \frac{g}{l} - \frac{a^2 k}{ml^2} & -\frac{a^2 d}{ml^2} & \frac{a^2 k}{ml^2} & \frac{a^2 d}{ml^2} \\ 0 & 0 & 0 & 1 \\ \frac{a^2 k}{ml^2} & \frac{a^2 d}{ml^2} & \frac{g}{l} - \frac{a^2 k}{ml^2} & -\frac{a^2 d}{ml^2} \end{bmatrix} \mathbf{x}(t) + \begin{bmatrix} 0 \\ \frac{1}{ml^2} \\ 0 \\ 0 \end{bmatrix} u(t) \\ y(t) &= [1 \quad 0 \quad 0 \quad 0] \mathbf{x}(t) + 0 \cdot u(t) \end{aligned}$$

```
In [1]: import numpy as np
        from numpy import sin, cos, tan, arcsin
```

```
In [2]: def EquilPoint(theta2, alpha):
        a = 3
        k= 1.5
        m = 2
        g = 9.81
        l = 6
        d = 1
        beta = 0.1
        temp = (-m*g*l*sin(theta2)-alpha+a**2*k*cos(theta2)*sin(theta2))/(a**2*k*cos(theta2))
        theta1 = arcsin(temp)
        T = -m*g*l*sin(theta1)-a**2*k*cos(theta1)*(sin(theta2)-sin(theta1))
        return theta1, T

        # For example for the fixed values of theta2 = 0.1 and alpha = 0.5
        theta1_bar, T_bar = EquilPoint(theta2=0.1, alpha=0.5)
        print(theta1_bar)
        print(T_bar)
```

```
-0.948098425224013
88.44306254140133
```

In [3]: `import sympy as sym`

```
def LinearizeModel(theta2, alpha):

    a = 3
    k = 1.5
    m = 2
    g = 9.81
    l = 6
    d = 1
    beta = 0.1
    theta1dot = 0
    theta2dot = 0

    theta1, T = EquilPoint(theta2, alpha) # calling the above EquilPoint function

    x1, x2, x3, x4, u1, u2 = sym.symbols('x1 x2 x3 x4 u1 u2')
    f = sym.Matrix([x2, (m*g*l*sym.sin(x1)+a**2*sym.cos(x1)*(k*(sym.sin(x3)-sym.sin(x1))+d*(x4-x2))+u1)/(m*l**2),
                    x4, (m*g*l*sym.sin(x3)-a**2*sym.cos(x3)*(k*(sym.sin(x3)-sym.sin(x1))+d*(x4-x2))+u2+beta*x4**2)/(m*l**2)])
    Asym = f.jacobian([x1, x2, x3, x4, u1, u2])
    Asub = Asym.subs([(x1, theta1), (x2, theta1dot), (x3, theta2), (x4, theta2dot), (u1, T), (u2, alpha)])

    Bsym = f.jacobian([u1, u2])
    Bsub = Bsym.subs([(x1, theta1), (x2, theta1dot), (x3, theta2), (x4, theta2dot), (u1, T), (u2, alpha)])

    C = np.array([1, 0, 0, 0])
    D = np.array([0, 0])

    return Asub, Bsub, C, D, theta1_bar, T_bar

# Calling the function
# for example consider the values of theta2_bar and alpha_bar to be 0.1 and 0.5, respectively
theta2_bar = 0.1
alpha_bar = 0.5
A, B, C, D, theta1_bar, T_bar = LinearizeModel(theta2_bar, alpha_bar)
print('A = ', A)
print('B = ', B)
print('C = ', C)
print('D = ', D)
print('theta1_bar = ', theta1_bar)
print('T_bar = ', T_bar)
```

```
A = Matrix([[0, 1, 0, 0, 0, 0], [1.02872606084634, -0.0729036009073981, 0.108809079849942, 0.0729036009073981, 1/72, 0], [0, 0, 0, 1, 0, 0], [0.108809079849942, 0.124375520659753, 1.45827472846828, -0.124375520659753, 0, 1/72]])
B = Matrix([[0, 0], [1/72, 0], [0, 0], [0, 1/72]])
C = [1 0 0 0]
D = [0 0]
theta1_bar = -0.948098425224013
T_bar = 88.44306254140133
```

---

## Question 2. Euler Discretization of a Building Heat Transfer Model

### Part (a)

Starting with our original equation:

$$m_z c_z \dot{T} = q + c_p u_1 (u_2 - T)$$

Use the Euler discretization approximation:

$$\dot{T} = \frac{T(k+1) - T(k)}{T_s}$$

Substitute in:

$$m_z c_z \frac{T(k+1) - T(k)}{T_s} = q(k) + c_p u_1(k) (u_2(k) - T(k))$$

Reshuffle variables:

$$T(k+1) = \left[ 1 - \frac{c_p T_s}{m_z c_z} u_1(k) \right] T(k) + \frac{T_s}{m_z c_z} [q(k) + c_p u_1(k) u_2(k)]$$

```
In [4]: def bldgHTM(T, u1, u2, q, mz, cz, cp):  
        Tdot = (q+cp*u1*(u2-T))/(mz*cz)  
        return Tdot
```

### Part (b)

```
In [5]: def eulerDiscretization(T,q,u1,u2):  
        mz = 100  
        cz = 20  
        Ts = 0.1 # Discretization sampling time  
        cp = 1000  
        T_KplusOne = (1-cp*Ts/(mz*cz)*u1)*T+Ts/(mz*cz)*(q+cp*u1*u2)  
        return T_KplusOne
```

---

## Question 3. Simulation of Nonlinear Bicycle Dynamics

### Part (a)

```
In [6]: import matplotlib.pyplot as plt

def carModel(beta, a, x, y, psi, v):

    l_r = 1.738
    dt = 0.1

    x_dot = v*cos(psi+beta)
    y_dot = v*sin(psi+beta)
    psi_dot = (v/l_r)*sin(beta)
    v_dot = a

    x_out = x + x_dot*dt
    y_out = y + y_dot*dt
    psi_out = psi + psi_dot*dt
    v_out = v + v_dot*dt

    return x_out, y_out, psi_out, v_out

# Calling the function example
x, y, psi, v = carModel(0.1,2,5,2,10,0.1)
print(x)
print(y)
print(psi)
print(v)
```

```
4.992194318198308
1.9937492935110712
10.000574415515805
0.30000000000000004
```

**Part (b)**

```

In [7]: from scipy.io import loadmat
Data = loadmat('sineData.mat')
a = Data['a']
beta = Data['beta']
time = Data['time']
Ts = 0.1

def sim(time, a, beta, x0):

    numSteps = len(time)

    x = x0[0]
    y = x0[1]
    psi = x0[2]
    v = x0[3]

    # Initialize trends
    xtrend = []
    xtrend.append(x)
    ytrend = []
    ytrend.append(y)
    psitrend = []
    psitrend.append(psi)
    vtrend = []
    vtrend.append(v)

    for i in range(0, numSteps-1):

        x, y, psi, v = carModel(beta[i], a[i], x, y, psi, v)
        xtrend.append(x)
        ytrend.append(y)
        psitrend.append(psi)
        vtrend.append(v)

    return np.asarray(xtrend, dtype=object), np.asarray(ytrend, dtype=object),
np.asarray(psitrend, dtype=object), np.asarray(vtrend, dtype=object)

# calling the function
x = 0.0
y = 0.0
psi = 0.0
v = 0.0
numSteps = len(time)

x0 = np.array([x, y, psi, v])
xtrend, ytrend, psitrend, vtrend = sim(time, a, beta, x0)

```

with given sequences of **time** and inputs **a** and **beta**. **time** is a vector of sampled time instants with sampling  $T_s$  starting at 0 seconds. You should use the function you write in part a. The input arguments (**time**, **a**, **beta**) should be 1001-by-1 vectors. **x0** should be a 4-by-1 vector. The output arguments should be all 1002-by-1 vectors.

### Part (c)

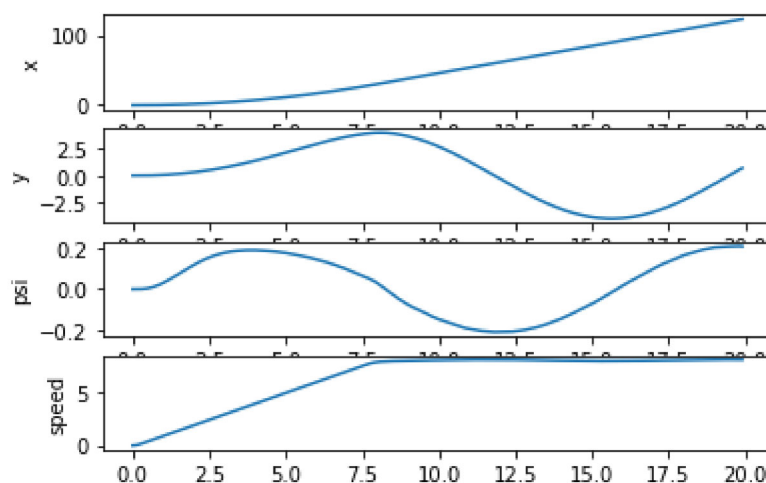
Write a script to test your code by using the file **sineData.mat**. This file contains three vectors: **time**, and the corresponding inputs at each time instant: **a** (acceleration) and **beta** ( $\beta$ ). Make a plot of the states versus time and  $y$  versus  $x$ . Create a simple animation that follows the trajectory. Use a bicycle or car shape that makes sense. Make sure the bike is pointing in the correct direction at every point.

```
In [8]: dt = time[1]-time[0]

# Shorten simulation time
numSteps = 200
time_red = time[0:numSteps]
xtrend_red = xtrend[0:numSteps]
ytrend_red = ytrend[0:numSteps]
psitrend_red = psitrend[0:numSteps]
vtrend_red = vtrend[0:numSteps]

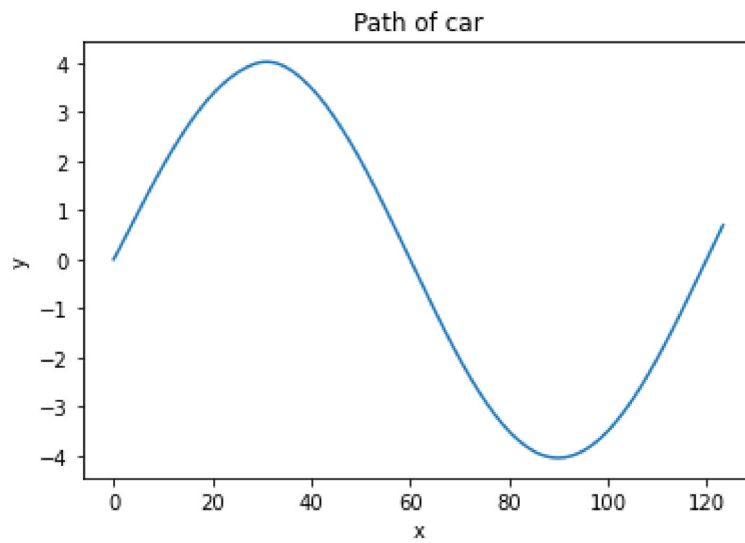
# Plot the results

plt.subplot(4,1,1)
plt.plot(time_red, xtrend_red)
plt.ylabel('x')
plt.subplot(4,1,2)
plt.plot(time_red, ytrend_red)
plt.ylabel('y')
plt.subplot(4,1,3)
plt.plot(time_red, psitrend_red)
plt.ylabel('psi')
plt.subplot(4,1,4)
plt.plot(time_red, vtrend_red)
plt.ylabel('speed')
plt.show()
```





```
In [9]: # Plot the path
plt.plot(xtrend_red,ytrend_red)
plt.xlabel('x')
plt.ylabel('y')
plt.title('Path of car')
plt.show()
```



```

In [10]: # Plot animation
from matplotlib import animation, rc
from IPython.display import HTML

w = 1
h = 2

fig = plt.figure(figsize=(10, 10))
ax = fig.add_subplot(111, autoscale_on=False, xlim=(np.min(xtrend_red)-5, np.m
ax(xtrend_red)[0]+5), ylim=(np.min(ytrend_red)[0]-5, np.max(ytrend_red)[0]+5))
ax.grid()

line, = ax.plot([], [], 'o-', lw=2)

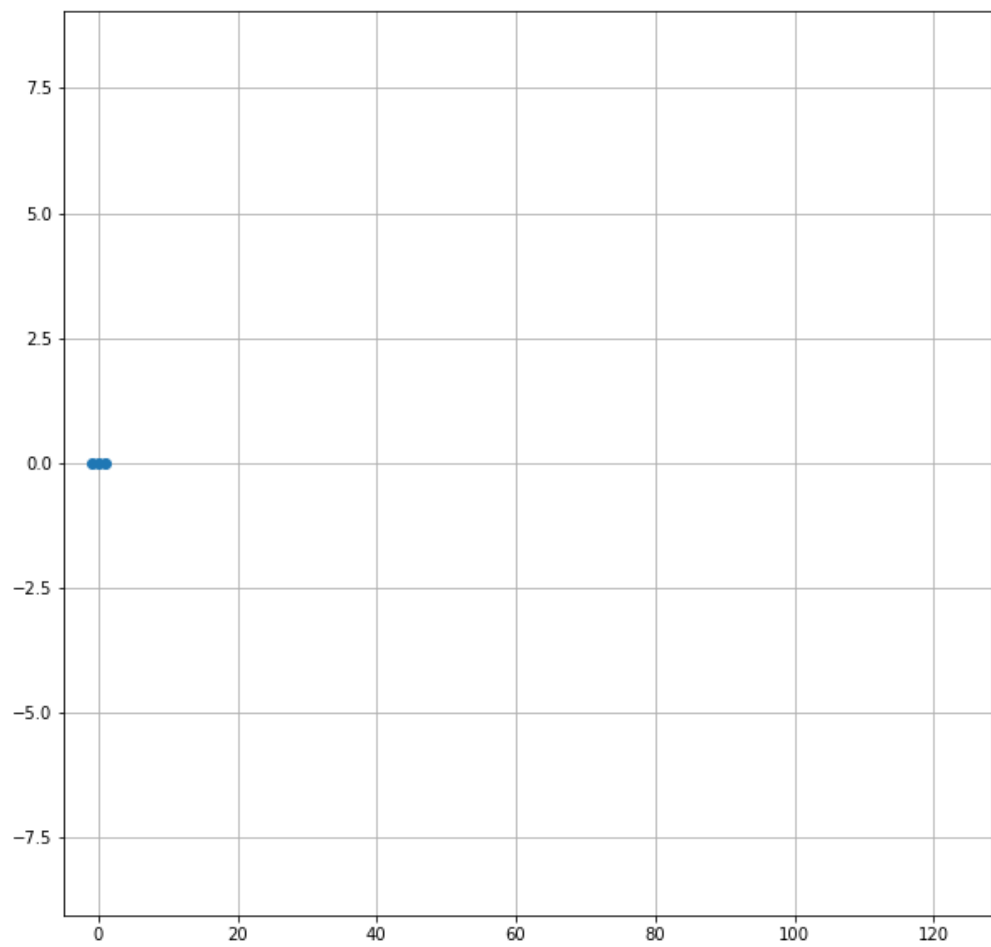
def init():
    line.set_data([], [])
    return line,

def animate(i):
    x = xtrend_red[i]
    y = ytrend_red[i]
    psi = psitrend_red[i]
    bikex = [x+cos(psi), x, x-cos(psi)]
    bikey = [y+sin(psi), y, y-sin(psi)]
    line.set_data(bikex, bikey)
    return line,

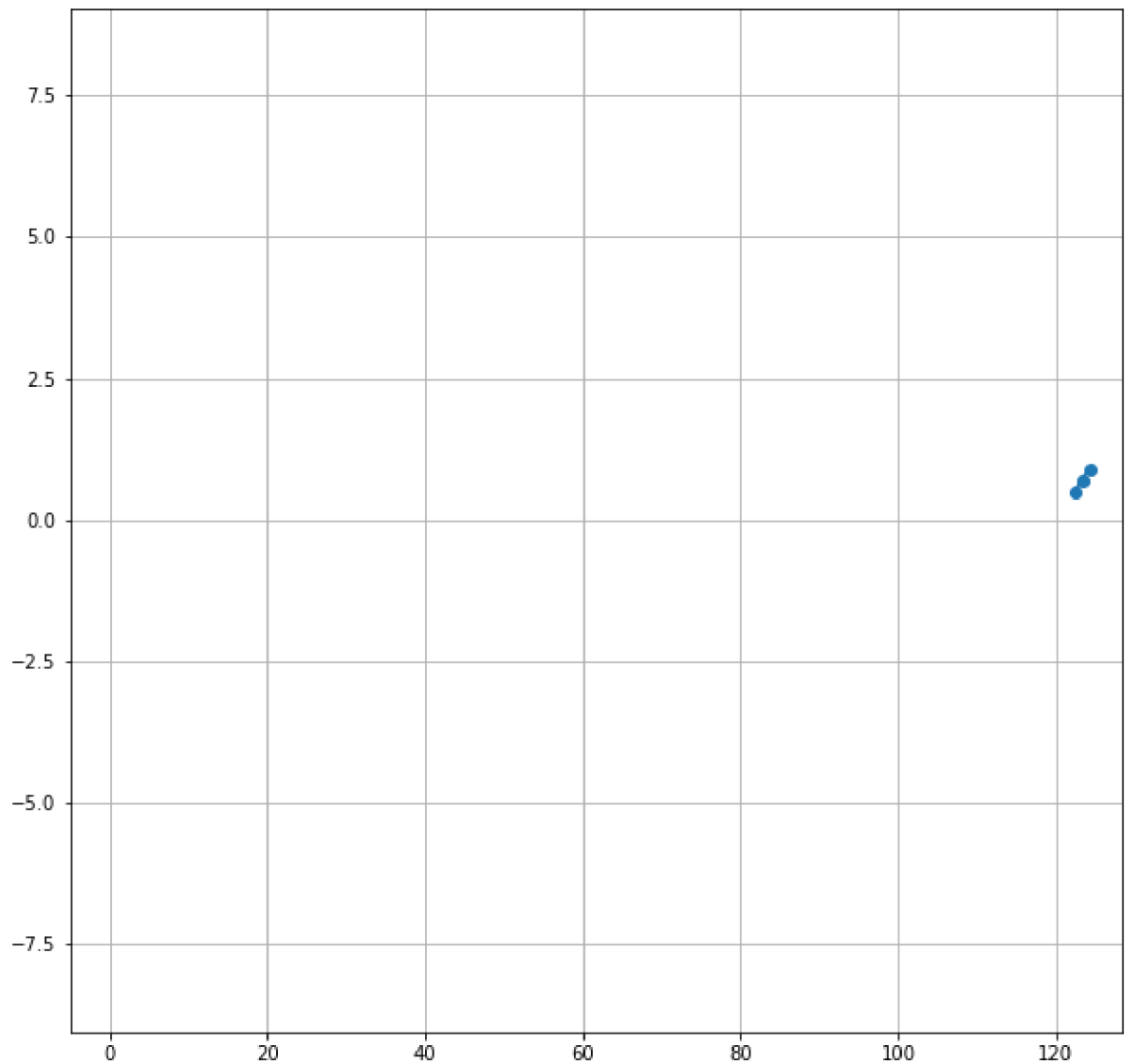
ani = animation.FuncAnimation(fig, animate, range(1, numSteps),
                              interval=0.1*1000, blit=True, init_func=init)
rc('animation', html='jshtml')
ani

```

Out[10]:



☐ Once ☒ Loop ☐ Reflect



## Question 4. Analysis of LTI Discrete-Time Systems

1. Let  $\alpha = 0$ . Is the system stable? - No, one of the eigenvalues is outside the unit circle ( $\lambda_{1,2,3,4} = \frac{1}{3}, \frac{1}{3}, -\frac{1}{2}, -\frac{5}{4}$ )
2. Now let  $\alpha = \frac{1}{2}$ . Is the system stable? No, again there is an eigenvalue outside the unit circle ( $\lambda_{1,2,3,4} = \frac{1}{3}, \frac{1}{3}, -\frac{3}{2}, -\frac{1}{4}$ )

```
In [11]: import scipy.linalg
def isSystemStable(alpha):

    A = np.array([[1/3, 0, 0, 0],
                  [0, -1/2, alpha, 0],
                  [0, 1/2, -5/4, 0],
                  [-1/2, 0, 0, 1/3]])

    eigVals, eigVecs = scipy.linalg.eig(A)
    TF = np.all(abs(eigVals)<1)

    return TF, eigVals
```

```
In [12]: alpha = 1
TF = isSystemStable(alpha)
TF
```

```
Out[12]: (False,
          array([-1.67539053+0.j, -0.07460947+0.j,  0.33333333+0.j,  0.33333333+0.j]))
```

```
In [13]: alpha = -0.5
TF = isSystemStable(alpha)
TF
```

```
Out[13]: (True,
          array([-0.875      +0.33071891j, -0.875      -0.33071891j,
                 0.33333333+0.j          ,  0.33333333+0.j          ]))
```