

die-2d

December 12, 2024

1 David Khachatryan: 2D Simulation of a Jack in a shaking box

1.0.1 Definition of coordinate frames

```
[632]: import sympy as sym
import numpy as np
import matplotlib.pyplot as plt
from sympy import sin, cos, Matrix, simplify, lambdify, Eq

[633]: # Define helper for generating SE(3) matrix
t = sym.symbols('t')

# rotational
def Rotz(theta):
    """Generates a 4x4 SE(3) transformation matrix for rotation about the
    ↪z-axis."""
    return sym.Matrix([
        # 4x4 rotation matrix
        [sym.cos(theta), -sym.sin(theta), 0, 0],
        [sym.sin(theta), sym.cos(theta), 0, 0],
        [0, 0, 1, 0],
        [0, 0, 0, 1]
    ])

# translational se(3)
def T(x, y):
    """Generates a 4x4 SE(3) transformation matrix for translation."""
    return sym.Matrix([
        # 4x4 translation matrix
        [1, 0, 0, x],
        [0, 1, 0, y],
        [0, 0, 1, 0],
        [0, 0, 0, 1]
    ])

# Define the helper for hatting a 3x1 vector
def hat(v: sym.Matrix) -> sym.Matrix:
    """
```

```

Converts a 3x1 vector into a 3x3 skew-symmetric matrix.

Parameters
-----
v : sympy.Matrix
    A 3x1 vector.

Returns
-----
sympy.Matrix
    A 3x3 skew-symmetric matrix.
"""
return sym.Matrix([[0, -v[2], v[1]],
                   [v[2], 0, -v[0]],
                   [-v[1], v[0], 0]])

# Define the helper for unhating a 3x3 skew-symmetric matrix
def unhat(v: sym.Matrix) -> sym.Matrix:
    """
    Extracts the 3x1 vector from a 3x3 skew-symmetric matrix.

    The unhat operation is the inverse of the hat operation.

    Parameters
    -----
    v : sympy.Matrix
        A 3x3 skew-symmetric matrix.

    Returns
    -----
    sympy.Matrix
        A 3x1 vector extracted from the skew-symmetric matrix.
    """
    V = Matrix([0, 0, 0, 0, 0, 0])
    V[0, 0] = v[0, 3]
    V[1, 0] = v[1, 3]
    V[2, 0] = v[2, 3]
    V[3, 0] = v[2, 1]
    V[4, 0] = v[0, 2]
    V[5, 0] = v[1, 0]
    return V

# Define the helper for getting the inverse of a g matrix
def inverse_g(g: sym.Matrix) -> sym.Matrix:
    """
    Computes the inverse of an SE(3) transformation matrix.

```

The inverse of an SE(3) matrix is given as:

$$g_{inv} = \begin{bmatrix} R.T & -R.T * p \\ 0 & 1 \end{bmatrix},$$

where:

- $R.T$ is the transpose of the rotation matrix.
- p is the position vector.

"""

$R = g[:3, :3]$

$p = g[:3, 3]$

return sym.Matrix([[R.T, -R.T * p], [0, 0, 0, 1]])

def g_dot_SE3(transform, angular_velocity):

"""

Computes the time derivative of an SE(3) transformation matrix.

The SE(3) matrix `transform` is composed of a rotation matrix `R` and a position vector `p`.

The derivative is given as:

$$g_{dot} = \begin{bmatrix} R_{dot} & p_{dot} \\ 0 & 0 \ 1 \end{bmatrix},$$

where:

- $R_{dot} = R * \hat{w}$,
 \hat{w} is the angular velocity vector.
- p_{dot} = derivative of the position vector `p`.

Parameters

`transform` : sympy.Matrix

The SE(3) homogeneous transformation matrix (3x3 in planar case).

`angular_velocity` : sympy.Symbol

Angular velocity (derivative of the rotation angle, θ).

Returns

sympy.Matrix

Time derivative of the SE(3) homogeneous transformation matrix (g_{dot}).

Notes

This assumes a 2D planar system with rotation about the z-axis (right-hand rule). The system uses:

- A 3x3 matrix for rotation and translation.
- Angular velocity vector $w = [0, 0, d(\theta)/dt]$, where z is the axis of rotation.

```

"""
# Compute the angular velocity vector (in matrix form for SE(3))
w = sym.Matrix([0, 0, angular_velocity.diff(t)])

# Extract Components
R = transform[:3, :3] # Rotation matrix
p = transform[:3, 3] # Translation vector

# Derivative of rotation: R_dot = R * hat(w)
R_dot = R * hat(w)

# Derivative of translation: p_dot
p_dot = p.diff(t)

# Construct g_dot
# use T and Rotz to construct the g_dot matrix
g_dot_matrix = sym.Matrix([[R_dot, p_dot], [0, 0, 0, 1]])
return g_dot_matrix

def body_velocity(transform, angular_velocity):
    """
    Computes the body velocity from the SE(3) transformation matrix.

    The body velocity is given as a 6D spatial velocity vector (in planar_
    ↪motion, reduced to 3D):
        V_body = [v ] -> Linear velocity from R_inv * (p_dot)
        [w ] -> Angular velocity

    Parameters
    -----
    transform : sympy.Matrix
        The SE(3) homogeneous transformation matrix for a frame (3x3 in planar_
    ↪case).
    angular_velocity : sympy.Symbol
        Angular velocity (derivative of the rotation angle, theta).

    Returns
    -----
    sympy.Matrix
        The spatial body velocity vector for the specified frame.

    Raises
    -----
    ValueError
        If an invalid output type is specified.

    Notes

```

```

-----
- Body velocity is computed as:
     $V_{\text{body}} = g^{-1} * g_{\text{dot}}$ 
    where  $g^{-1}$  is the inverse of the SE(3) matrix.
- This function is planar, so angular velocity is scalar (about z-axis).
"""

# Compute  $g^{-1}$  (the inverse of SE(3))
transform_inverse = inverse_g(transform)

# Compute  $g_{\text{dot}}$ 
g_dot_matrix = g_dot_SE3(transform, angular_velocity)

# Multiply to get body velocity:  $V_{\text{body}} = g^{-1} * g_{\text{dot}}$ 
V_body_matrix = transform_inverse * g_dot_matrix

# Extract body velocity:  $w$  = angular component,  $v$  = linear component
w_hat = V_body_matrix[:3, :3]
v = V_body_matrix[:3, 3]
w = unhat(w_hat)
return v.col_join(w)

def integrate(f, xt, dt, time):
    """
    This function takes in an initial condition  $x(t)$  and a timestep  $dt$ ,
    as well as a dynamical system  $f(x)$  that outputs a vector of the
    same dimension as  $x(t)$ . It outputs a vector  $x(t+dt)$  at the future
    time step.

    Parameters
    =====
    dyn: Python function
        derivate of the system at a given step  $x(t)$ ,
        it can considered as  $\dot{x}(t) = \text{func}(x(t))$ 
    xt: NumPy array
        current step  $x(t)$ 
    dt:
        step size for integration

    Return
    =====
    new_xt:
        value of  $x(t+dt)$  integrated from  $x(t)$ 
    """
    k1 = dt * f(xt, time)
    k2 = dt * f(xt+k1/2., time)
    k3 = dt * f(xt+k2/2., time)
    k4 = dt * f(xt+k3, time)

```

```

new_xt = xt + (1/6.) * (k1+2.0*k2+2.0*k3+k4)
return new_xt

```

<>:187: SyntaxWarning:

invalid escape sequence '\d'

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invalid escape sequence '\d'

/tmp/ipykernel_984525/3401856478.py:187: SyntaxWarning:

invalid escape sequence '\d'

```

[634]: # We have 6 configuration variables
x_b = sym.Function('x_b')(t)
y_b = sym.Function('y_b')(t)
theta_b = sym.Function('theta_b')(t)
x_j = sym.Function('x_j')(t)
y_j = sym.Function('y_j')(t)
theta_j = sym.Function('theta_j')(t)
lamb = sym.symbols(r'lambda')
x_b_dot_Plus, y_b_dot_Plus, theta_b_dot_Plus, x_j_dot_Plus, y_j_dot_Plus,
    ↪theta_j_dot_Plus = sym.symbols(r'x_b_dot_+, y_b_dot_+, theta_b_dot_+,
    ↪x_j_dot_+, y_j_dot_+, theta_j_dot_+')
xbl, ybl, tbl, xjl, yjl, tjl, xbl_dot, ybl_dot, tbl_dot, xjldot, yjldot, tjldot =
    ↪sym.symbols('x_box_l, y_box_l, theta_box_l, x_jack_l, y_jack_l,
    ↪theta_jack_l, x_box_ldot, y_box_ldot, theta_box_ldot, x_jack_ldot,
    ↪y_jack_ldot, theta_jack_ldot')

q = Matrix([
    x_b,
    y_b,
    theta_b,
    x_j, y_j,
    theta_j]
)
qdot = q.diff(t)
qddot = qdot.diff(t)

```

```

[635]: # Parameters for the box and jack
box_length, box_mass = 4, 50 # Box length and mass
box_moi = (4) * box_mass * box_length ** 2 # Moment of inertia for the box

jack_length, m_jack = 1, 1 # Jack length and mass

```

```

jack_moi = (4) * m_jack * (jack_length) ** 2 # Moment of inertia for the jack
jack_mass = 1 # Mass of the jack
g = 9.81

# Homogeneous transformation matrices
g_wa = sym.Matrix([
    [cos(theta_b), -sin(theta_b), 0, x_b],
    [sin(theta_b), cos(theta_b), 0, y_b],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_wb = sym.Matrix([
    [cos(theta_j), -sin(theta_j), 0, x_j],
    [sin(theta_j), cos(theta_j), 0, y_j],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_a_a1 = sym.Matrix([
    [1, 0, 0, box_length],
    [0, 1, 0, box_length],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_a_a2 = sym.Matrix([
    [1, 0, 0, 0],
    [0, 1, 0, -box_length],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_a_a3 = sym.Matrix([
    [1, 0, 0, -box_length],
    [0, 1, 0, 0],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_a_a4 = sym.Matrix([
    [1, 0, 0, 0],
    [0, 1, 0, box_length],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

```

```

g_b_b1 = sym.Matrix([
    [1, 0, 0, jack_length],
    [0, 1, 0, 0],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_b_b2 = sym.Matrix([
    [1, 0, 0, 0],
    [0, 1, 0, -jack_length],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_b_b3 = sym.Matrix([
    [1, 0, 0, -jack_length],
    [0, 1, 0, 0],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_b_b4 = sym.Matrix([
    [1, 0, 0, 0],
    [0, 1, 0, jack_length],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
])

g_w_a1 = g_wa @ g_a_a1
g_w_a2 = g_wa @ g_a_a2
g_w_a3 = g_wa @ g_a_a3
g_w_a4 = g_wa @ g_a_a4

g_w_j1 = g_wb @ g_b_b1
g_w_j2 = g_wb @ g_b_b2
g_w_j3 = g_wb @ g_b_b3
g_w_j4 = g_wb @ g_b_b4

g_a1_j1 = inverse_g(g_w_a1) @ g_w_j1
g_a1_j2 = inverse_g(g_w_a1) @ g_w_j2
g_a1_j3 = inverse_g(g_w_a1) @ g_w_j3
g_a1_j4 = inverse_g(g_w_a1) @ g_w_j4

g_a2_j1 = inverse_g(g_w_a2) @ g_w_j1
g_a2_j2 = inverse_g(g_w_a2) @ g_w_j2
g_a2_j3 = inverse_g(g_w_a2) @ g_w_j3
g_a2_j4 = inverse_g(g_w_a2) @ g_w_j4

```



```

g_a3_j1 = inverse_g(g_w_a3) @ g_w_j1
g_a3_j2 = inverse_g(g_w_a3) @ g_w_j2
g_a3_j3 = inverse_g(g_w_a3) @ g_w_j3
g_a3_j4 = inverse_g(g_w_a3) @ g_w_j4

g_a4_j1 = inverse_g(g_w_a4) @ g_w_j1
g_a4_j2 = inverse_g(g_w_a4) @ g_w_j2
g_a4_j3 = inverse_g(g_w_a4) @ g_w_j3
g_a4_j4 = inverse_g(g_w_a4) @ g_w_j4

```

```

[636]: origin = sym.Matrix([0, 0, 0, 1])
r_wa = g_wa @ origin
r_wb = g_wb @ origin

```

```

[637]: # Now calculate velocities of the box and jack
#v_a = sym.simplify(body_velocity(g_wa, theta_b))
#v_b = sym.simplify(body_velocity(g_wb, theta_j))
#
v_a = unhat(inverse_g(g_wa) @ g_wa.diff(t))
v_b = unhat(inverse_g(g_wb) @ g_wb.diff(t))

```

```

[638]: # Now calculate inertia
I_a = sym.Matrix([
    [4*box_mass, 0, 0, 0, 0, 0],
    [0, 4*box_mass, 0, 0, 0, 0],
    [0, 0, 4*box_mass, 0, 0, 0],
    [0, 0, 0, 0, 0, 0],
    [0, 0, 0, 0, 0, 0],
    [0, 0, 0, 0, 0, box_moi]
])

I_b = sym.Matrix([
    [4*jack_mass, 0, 0, 0, 0, 0],
    [0, 4*jack_mass, 0, 0, 0, 0],
    [0, 0, 4*jack_mass, 0, 0, 0],
    [0, 0, 0, 0, 0, 0],
    [0, 0, 0, 0, 0, 0],
    [0, 0, 0, 0, 0, jack_moi]
])

```

1.1 Euler-Lagrange equations

```

[639]: # Now calculate the kinetic energy
KE = sym.simplify((1/2)*v_a.T*I_a*v_a + (1/2)*v_b.T*I_b*v_b)[0]
V = 4 * box_mass * g * y_b + 4 * jack_mass * g * y_j
L = KE - V

```

```

## Now calculate the Euler-Lagrange equations
#L_qdot = L.diff(qdot)
#L_qdot_dot = L_qdot.diff(t)
#L_q = L.diff(q)
#eqs = sym.simplify(L_q - L_qdot_dot)
#
#lhs = eqs
#rhs = F_ext
#EL_eqs = Eq(rhs, lhs)
#display(EL_eqs)
#
## Now solve the equations
#sol = sym.solve(EL_eqs, qddot, dict=True)
#display(sol)
# External forces (box shaking parameters)
k = 10000
theta_d_b = sin(np.pi * t / 2.5)
F_theta_b = k * theta_d_b
F_y_b = 4 * box_mass * g
F_ext = Matrix([0, F_y_b, F_theta_b, 0, 0, 0])

# Solve lagrange equations
dL_dq = simplify(Matrix([L]).jacobian(q).T)
dL_dqdot = simplify(Matrix([L]).jacobian(qdot).T)
ddL_dqdot_dt = simplify(dL_dqdot.diff(t))

lhs = simplify(ddL_dqdot_dt - dL_dq)
rhs = simplify(F_ext)
EL_Eqs = simplify(Eq(lhs, rhs))
# Solve the Euler-Lagrange Equations:
sol = sym.solve(EL_Eqs, qddot, dict=True)

# Compute the Hamiltonian:
H = simplify((dL_dqdot.T * qdot)[0] - L)

```

```

[640]: # Create a dictionary to store the lambdified functions
ddot_funcs = {
    'x_box': lambdify([*q, *qdot, t], sol[0][qddot[0]]),
    'y_box': lambdify([*q, *qdot, t], sol[0][qddot[1]]),
    'theta_box': lambdify([*q, *qdot, t], sol[0][qddot[2]]),
    'x_jack': lambdify([*q, *qdot, t], sol[0][qddot[3]]),
    'y_jack': lambdify([*q, *qdot, t], sol[0][qddot[4]]),
    'theta_jack': lambdify([*q, *qdot, t], sol[0][qddot[5]])
}

def dynamics(s, t):
    # Compute accelerations using the lambdified functions

```

```

q_ddots = [func(*s, t) for func in ddot_funcs.values()]

# Combine velocity and acceleration into the state derivative
sdot = np.array([
    *s[6:], # Velocities
    *q_ddots # Accelerations
])

return sdot

```

```

[641]: #$ Tau plus handling boilerplate:
elements = []
for i in range(6):
    elements.extend([qdot[i], sol[0][qddot[i]]])

qddot_Matrix = Matrix(elements)
state_variable_mapping = {
    q[i]: vars()[f'{var}_l'] for i, var in enumerate(['x_b', 'y_b', 'theta_b',
    ↪ 'x_j', 'y_j', 'theta_j'])
}
state_variable_mapping.update({
    qdot[i]: vars()[f'{var}_ldot'] for i, var in enumerate(['x_b', 'y_b',
    ↪ 'theta_b', 'x_j', 'y_j', 'theta_j'])
})

# Substitute the variables in the qddot matrix
qddot_d = qddot_Matrix.subs(state_variable_mapping)

# Define the lambdified function
qddot_lambdify = lambdify(
    [
        xbl, xbldot, ybl, ybldot, tbl, tbldot,
        xjl, xjldot, yjl, yjldot, tjl, tjldot, t
    ],
    qddot_d
)

```

1.2 Defining impact constraints

```

[ ]: # Wall impact handling boilerplate:
wall_b1_j1 = (g_a1_j1[3]).subs(state_variable_mapping)
wall_b1_j2 = (g_a1_j2[3]).subs(state_variable_mapping)
wall_b1_j3 = (g_a1_j3[3]).subs(state_variable_mapping)
wall_b1_j4 = (g_a1_j4[3]).subs(state_variable_mapping)
wall_b2_j1 = (g_a2_j1[7]).subs(state_variable_mapping)
wall_b2_j2 = (g_a2_j2[7]).subs(state_variable_mapping)
wall_b2_j3 = (g_a2_j3[7]).subs(state_variable_mapping)

```

```

wall_b2_j4 = (g_a2_j4[7]).subs(state_variable_mapping)
wall_b3_j1 = (g_a3_j1[3]).subs(state_variable_mapping)
wall_b3_j2 = (g_a3_j2[3]).subs(state_variable_mapping)
wall_b3_j3 = (g_a3_j3[3]).subs(state_variable_mapping)
wall_b3_j4 = (g_a3_j4[3]).subs(state_variable_mapping)
wall_b4_j1 = (g_a4_j1[7]).subs(state_variable_mapping)
wall_b4_j2 = (g_a4_j2[7]).subs(state_variable_mapping)
wall_b4_j3 = (g_a4_j3[7]).subs(state_variable_mapping)
wall_b4_j4 = (g_a4_j4[7]).subs(state_variable_mapping)

# Constraint
constraint = simplify(
    Matrix([
        [wall_b1_j1], [wall_b1_j2], [wall_b1_j3], [wall_b1_j4],
        [wall_b2_j1], [wall_b2_j2], [wall_b2_j3], [wall_b2_j4],
        [wall_b3_j1], [wall_b3_j2], [wall_b3_j3], [wall_b3_j4],
        [wall_b4_j1], [wall_b4_j2], [wall_b4_j3], [wall_b4_j4]
    ])
)

Hamiltonian_ = H.subs(state_variable_mapping)
dL_dqdot_dum = dL_dqdot.subs(state_variable_mapping)
dPhidq_dum = constraint.jacobian([xbl, ybl, tbl, xjl, yjl, tj1])
impact_dict = {
    xbl_dot:x_b_dot_Plus,
    ybl_dot:y_b_dot_Plus,
    tbl_dot:theta_b_dot_Plus,
    xjl_dot:x_j_dot_Plus,
    yjl_dot:y_j_dot_Plus,
    tj1_dot:theta_j_dot_Plus
}

# tau+ evaluations:
dL_dqdot_dumPlus = simplify(dL_dqdot_dum.subs(impact_dict))
dPhidq_dumPlus = simplify(dPhidq_dum.subs(impact_dict))
Hamiltonian_Plus = simplify(Hamiltonian_.subs(impact_dict))
impact_eqns_list = []

# Define equations
lhs = Matrix([dL_dqdot_dumPlus[0] - dL_dqdot_dum[0],
              dL_dqdot_dumPlus[1] - dL_dqdot_dum[1],
              dL_dqdot_dumPlus[2] - dL_dqdot_dum[2],
              dL_dqdot_dumPlus[3] - dL_dqdot_dum[3],
              dL_dqdot_dumPlus[4] - dL_dqdot_dum[4],
              dL_dqdot_dumPlus[5] - dL_dqdot_dum[5],
              Hamiltonian_Plus - Hamiltonian_])

```

```

for i in range(constraint.shape[0]):
    rhs = Matrix([lamb*dPhidq_dum[i,0],
                  lamb*dPhidq_dum[i,1],
                  lamb*dPhidq_dum[i,2],
                  lamb*dPhidq_dum[i,3],
                  lamb*dPhidq_dum[i,4],
                  lamb*dPhidq_dum[i,5],
                  0])
    impact_eqns_list.append(simplify(Eq(lhs, rhs)))

```

1.3 Impact Update

```

[ ]: dum_list = [
    x_b_dot_Plus, y_b_dot_Plus, theta_b_dot_Plus,
    x_j_dot_Plus, y_j_dot_Plus, theta_j_dot_Plus
]

def impact_update(s, impact_eqns, dum_list):
    """
    This function takes in the current state of the system and the impact_
    equations
    and returns the updated state of the system after the impact.

    Parameters
    =====
    s: NumPy array
        current state of the system

    impact_eqns: list
        list of impact equations

    dum_list: list
        list of dummy variables

    Return
    =====
    s: NumPy array
        updated state of the system
    """
    subs = {
        xbl:s[0], ybl:s[1], tbl:s[2],
        xjl:s[3], yjl:s[4], tj1:s[5],
        xbl_dot:s[6], ybl_dot:s[7], tbladot:s[8],
        xjldot:s[9], yjldot:s[10], tjldot:s[11]
    }
    new_impact_eqns = impact_eqns.subs(subs)
    impact_solns = sym.solve(

```

```

        new_impact_eqns,
        [
            x_b_dot_Plus, y_b_dot_Plus, theta_b_dot_Plus,
            x_j_dot_Plus, y_j_dot_Plus, theta_j_dot_Plus,
            lamb
        ],
        dict=True
    )

    if len(impact_solns) != 1:
        for sol in impact_solns:
            lamb_sol = sol[lamb]
            if abs(lamb_sol) > 1e-06:
                return np.array([
                    *s[:6],
                    float(sym.N(sol[dum_list[0]])),
                    float(sym.N(sol[dum_list[1]])),
                    float(sym.N(sol[dum_list[2]])),
                    float(sym.N(sol[dum_list[3]])),
                    float(sym.N(sol[dum_list[4]])),
                    float(sym.N(sol[dum_list[5]])),
                ])

```

1.4 Impact Condition

```

[ ]: phi_func = lambdify(
    [
        xbl, ybl, tbl,
        xjl, yjl, tjl,
        xbl_dot, ybl_dot, tbl_dot,
        xjl_dot, yjl_dot, tjl_dot
    ],
    constraint
)

def impact_condition(s, phi_func, threshold = 1e-1):
    """
    This function checks if the system is in impact condition.
    It returns True if the system is in impact condition, False otherwise.

    Parameters
    -----
    s : np.array
        The state of the system.

    phi_func : function
        The function that calculates the impact constraints.

```

```

threshold : float
    The threshold for the impact condition.

Returns
-----
int: The index of the impact constraint that is less than the threshold.
    -1 if no impact condition is met.
"""

# Get the impact constraints
phi_val = phi_func(*s)

# Check if any of the constraints are less than the threshold
for i in range(phi_val.shape[0]):
    if abs(phi_val[i]) < threshold:
        return i

return -1

```

```

[645]: def simulate_impact(f, x0, tspan, dt, integrate):
    """
    This function simulates the trajectory of a dynamical system
    from a given initial condition x0, over a time span tspan,
    with a time step dt. It uses the numerical integration method
    specified in the input argument 'integrate'.

    Parameters
    =====
    f: Python function
        derivate of the system at a given step x(t),
        it can considered as  $\dot{x}(t) = \text{func}(x(t))$ 
    x0: NumPy array
        initial conditions
    tspan: Python list
        tspan = [min_time, max_time], it defines the start and end
        time of simulation
    dt:
        time step for numerical integration
    integrate: Python function
        numerical integration method used in this simulation

    Return
    =====
    x_traj:
        simulated trajectory of x(t) from t=0 to tf
    """

```

```

# Initialize the count of the number of time steps
num = int((max(tspan) - min(tspan)) / dt)

# Copy the initial condition
x = np.copy(x0)

# Initialize the trajectory array
xtraj = np.zeros((len(x0), num))
time = 0
for i in range(num):
    # Update the time
    time += dt

    # Check for impact condition
    impact = impact_condition(x, phi_func, 1e-1)
    if impact != -1:
        # Update the system after impact
        x = impact_update(x, impact_eqns_list[impact], dum_list)

    # Integrate the system
    xtraj[:, i] = integrate(f, x, dt, time)

    # Update the state for the next iteration
    x = np.copy(xtraj[:, i])
return xtraj

```

<>:2: SyntaxWarning:

invalid escape sequence '\d'

<>:2: SyntaxWarning:

invalid escape sequence '\d'

/tmp/ipykernel_984525/520826766.py:2: SyntaxWarning:

invalid escape sequence '\d'

```

[646]: # Simulate the motion:
tspan = [0, 10]
dt = 0.01
s0 = np.array([0, 0, 0, 0, 0, 0, 0, 0, 0, -2.2, 0, 0, 0])

N = int((max(tspan) - min(tspan))/dt)
tvec = np.linspace(min(tspan), max(tspan), N)
traj = simulate_impact(dynamics, s0, tspan, dt, integrate)

```



```

plt.figure()
plt.plot(tvec, traj[0], label='x_box')
plt.plot(tvec, traj[1], label='y_box')
plt.plot(tvec, traj[2], label='theta_box')
plt.title('Box Placement Simulation')
plt.xlabel('t')

# Add the plot labels as legends on the plot
plt.legend()

plt.show()

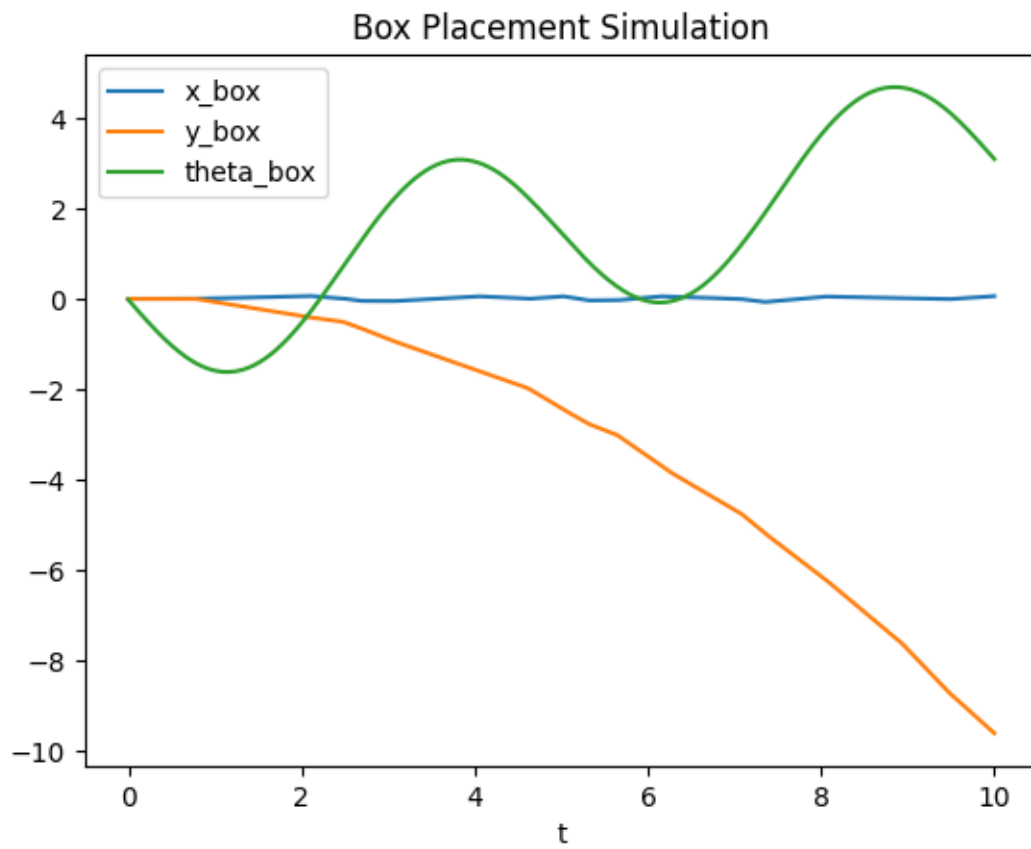
plt.figure()
plt.plot(tvec, traj[3], label='x_jack')
plt.plot(tvec, traj[4], label='y_jack')
plt.plot(tvec, traj[5], label='theta_jack')
plt.title('Jack Placement Simulation')
plt.xlabel('t')
plt.legend()
plt.show()

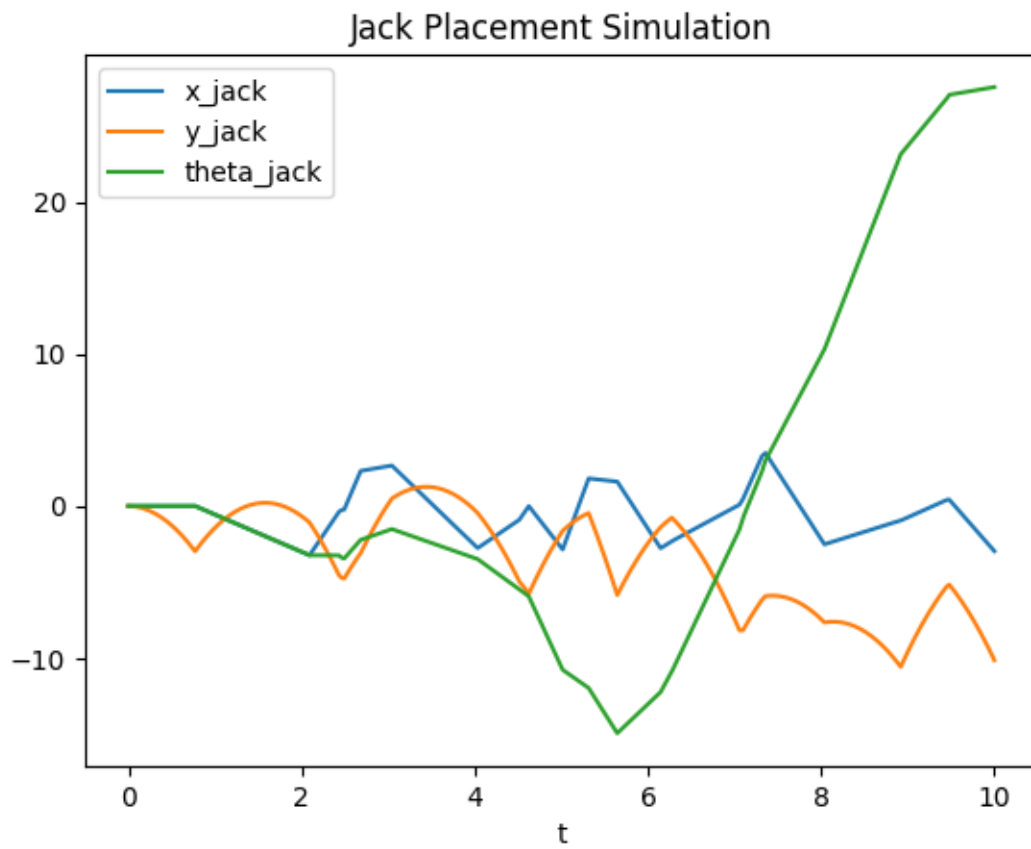
plt.figure()
plt.plot(tvec, traj[6], label='x_box_dot')
plt.plot(tvec, traj[7], label='y_box_dot')
plt.plot(tvec, traj[8], label='theta_box_dot')
plt.title('Box Velocity Simulation')
plt.xlabel('t')
plt.legend()
plt.show()

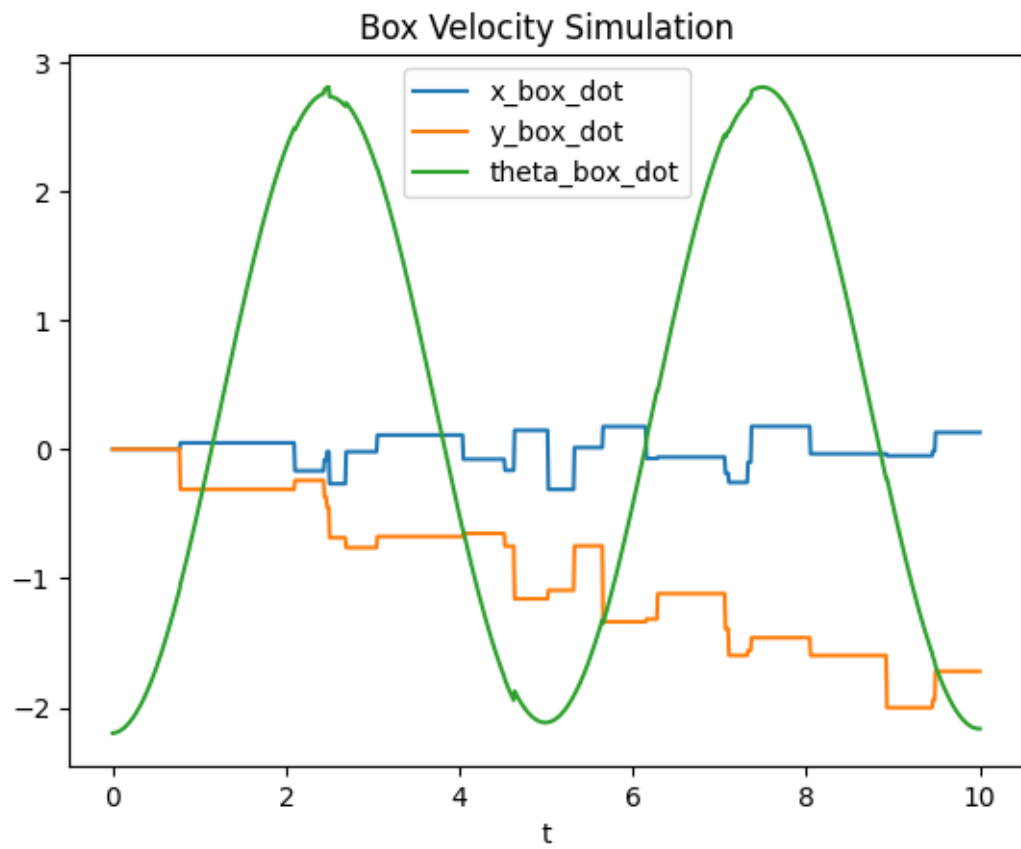
plt.figure()
plt.plot(tvec, traj[9], label='x_jack_dot')
plt.plot(tvec, traj[10], label='y_jack_dot')
plt.plot(tvec, traj[11], label='theta_jack_dot')
plt.title('Jack Velocity Simulation')
plt.legend()
plt.xlabel('t')

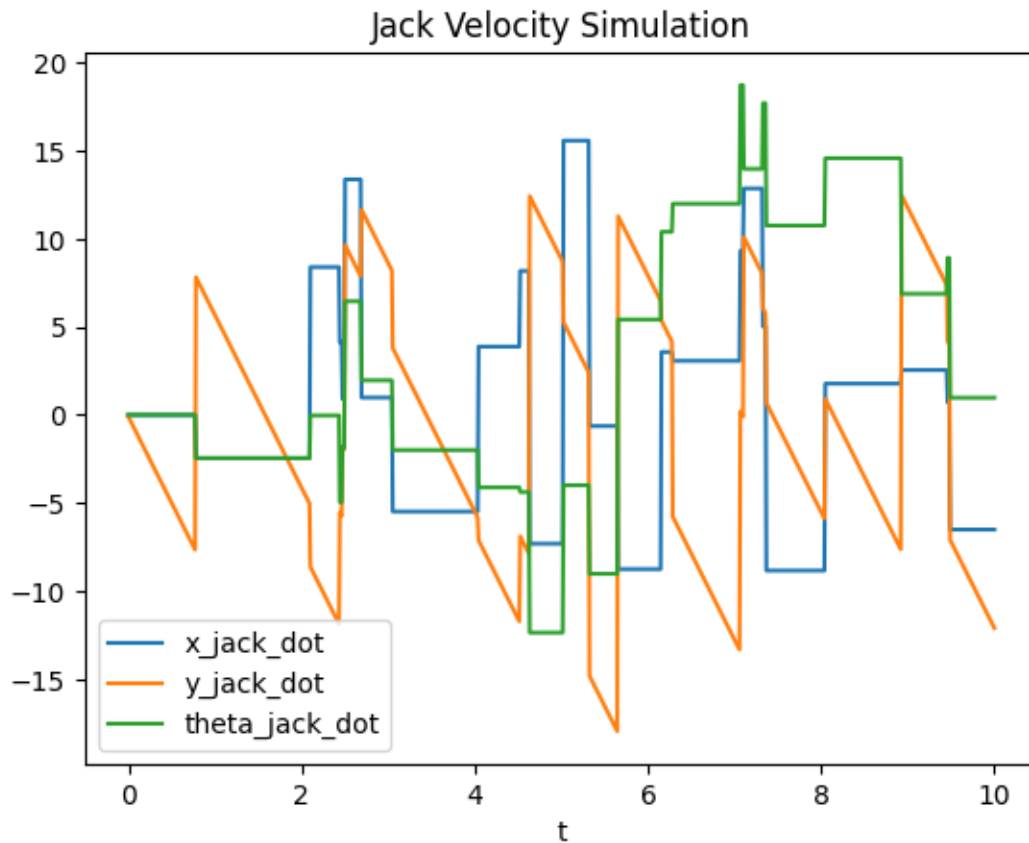
plt.show()

```









```
[647]: def animate(config_array,l=1,w=0.2,T=10):
        """
        Function to generate web-based animation of the system

        Parameters:
        =====
        config_array:
            trajectory of theta1 and theta2, should be a NumPy array with
            shape of (2,N)
        L1:
            length of the first pendulum
        L2:
            length of the second pendulum
        T:
            length/seconds of animation duration

        Returns: None
        """
        #####
```

```

# Imports required for animation. (leave this part)
from plotly.offline import init_notebook_mode, iplot
from IPython.display import display, HTML
import plotly.graph_objects as go

#####
# Browser configuration. (leave this part)
def configure_plotly_browser_state():
    import IPython
    display(IPython.core.display.HTML('''
        <script src="/static/components/requirejs/require.js"></script>
        <script>
            requirejs.config({
                paths: {
                    base: '/static/base',
                    plotly: 'https://cdn.plot.ly/plotly-1.5.1.min.js?noext',
                },
            });
        </script>
        '''))
configure_plotly_browser_state()
init_notebook_mode(connected=False)

#####
# Getting data from pendulum angle trajectories.
N = len(config_array[0])

x_box_array = config_array[0]
y_box_array = config_array[1]
theta_box_array = config_array[2]
x_jack_array = config_array[3]
y_jack_array = config_array[4]
theta_jack_array = config_array[5]

b1_x_array = np.zeros(N, dtype=np.float32)
b1_y_array = np.zeros(N, dtype=np.float32)
b2_x_array = np.zeros(N, dtype=np.float32)
b2_y_array = np.zeros(N, dtype=np.float32)
b3_x_array = np.zeros(N, dtype=np.float32)
b3_y_array = np.zeros(N, dtype=np.float32)
b4_x_array = np.zeros(N, dtype=np.float32)
b4_y_array = np.zeros(N, dtype=np.float32)

j_x_array = np.zeros(N, dtype=np.float32)
j_y_array = np.zeros(N, dtype=np.float32)
j1_x_array = np.zeros(N, dtype=np.float32)

```

```

j1_y_array = np.zeros(N, dtype=np.float32)
j2_x_array = np.zeros(N, dtype=np.float32)
j2_y_array = np.zeros(N, dtype=np.float32)
j3_x_array = np.zeros(N, dtype=np.float32)
j3_y_array = np.zeros(N, dtype=np.float32)
j4_x_array = np.zeros(N, dtype=np.float32)
j4_y_array = np.zeros(N, dtype=np.float32)

for t in range(N):
    g_w_b = np.array([[np.cos(theta_box_array[t]), -np.
↪sin(theta_box_array[t]), 0, x_box_array[t]],
                      [np.sin(theta_box_array[t]), np.
↪cos(theta_box_array[t]), 0, y_box_array[t]],
                      [0, 0, 1, 0],
                      [0, 0, 0, 1]])

    g_w_j = np.array([[np.cos(theta_jack_array[t]), -np.
↪sin(theta_jack_array[t]), 0, x_jack_array[t]],
                      [np.sin(theta_jack_array[t]), np.
↪cos(theta_jack_array[t]), 0, y_jack_array[t]],
                      [0, 0, 1, 0],
                      [0, 0, 0, 1]])

    b1 = g_w_b.dot(np.array([box_length, box_length, 0, 1]))
    b1_x_array[t] = b1[0]
    b1_y_array[t] = b1[1]
    b2 = g_w_b.dot(np.array([box_length, -box_length, 0, 1]))
    b2_x_array[t] = b2[0]
    b2_y_array[t] = b2[1]
    b3 = g_w_b.dot(np.array([-box_length, -box_length, 0, 1]))
    b3_x_array[t] = b3[0]
    b3_y_array[t] = b3[1]
    b4 = g_w_b.dot(np.array([-box_length, box_length, 0, 1]))
    b4_x_array[t] = b4[0]
    b4_y_array[t] = b4[1]

    j = g_w_j.dot(np.array([0, 0, 0, 1]))
    j_x_array[t] = j[0]
    j_y_array[t] = j[1]
    j1 = g_w_j.dot(np.array([jack_length, 0, 0, 1]))
    j1_x_array[t] = j1[0]
    j1_y_array[t] = j1[1]
    j2 = g_w_j.dot(np.array([0, -jack_length, 0, 1]))
    j2_x_array[t] = j2[0]
    j2_y_array[t] = j2[1]
    j3 = g_w_j.dot(np.array([-jack_length, 0, 0, 1]))

```

```

j3_x_array[t] = j3[0]
j3_y_array[t] = j3[1]
j4 = g_w_j.dot(np.array([0, jack_length, 0, 1]))
j4_x_array[t] = j4[0]
j4_y_array[t] = j4[1]

#####
# Axis limits.
xm = -11
xM = 11
ym = -15
yM = 11

#####
# Defining data dictionary.
data=[dict(name = 'Box'),
       dict(name = 'Jack'),
       dict(name = 'Mass1_Jack'),
]

#####
# Preparing simulation layout.
layout=dict(autosize=False, width=1000, height=1000,
            xaxis=dict(range=[xm, xM], autorange=False,
↪zeroline=True,dtick=1),
            yaxis=dict(range=[ym, yM], autorange=False,
↪zeroline=False,scaleanchor = "x",dtick=1),
            title='2D: Jack in a Box Simulation',
            hovermode='closest',
            updatemenus= [{ 'type': 'buttons',
                            'buttons': [{ 'label': 'Play','method': 'animate',
↪{'args': [None, {'frame':
↪{'args': [[None], {'frame':
↪{'duration': T, 'redraw': False}},
                            {'transition': {'duration':
↪0}}], 'label': 'Pause','method': 'animate'}
                            ]
                            }
                            ]
            })

#####
# Defining the frames of the simulation.
frames=[dict(data=[
↪dict(x=[b1_x_array[k],b2_x_array[k],b3_x_array[k],b4_x_array[k],b1_x_array[k]],

```



```

        ↪y=[b1_y_array[k],b2_y_array[k],b3_y_array[k],b4_y_array[k],b1_y_array[k]],
            mode='lines',
            line=dict(color='purple', width=3)
        ),

        ↪dict(x=[j1_x_array[k],j3_x_array[k],j_x_array[k],j2_x_array[k],j4_x_array[k]],
            ↪y=[j1_y_array[k],j3_y_array[k],j_y_array[k],j2_y_array[k],j4_y_array[k]],
                mode='lines',
                line=dict(color='black', width=3)
            ),
        go.Scatter(
            x=[j1_x_array[k],j2_x_array[k],j3_x_array[k],j4_x_array[k]],
            y=[j1_y_array[k],j2_y_array[k],j3_y_array[k],j4_y_array[k]],
            mode="markers",
            marker=dict(color='red', size=6)),
        ]) for k in range(N)]

#####
# Putting it all together and plotting.
figure1=dict(data=data, layout=layout, frames=frames)
iplot(figure1)

#####
# The animation:
animate(traj)

```

<IPython.core.display.HTML object>

[]:

[]:

[]: