

EECS106A Discussion 9: Lagrangian dynamics in 3D and state spaces

1 Review: Lagrangian dynamics

As we saw last week, the kinetic energy of a rigid body can be expressed as:

$$T = \frac{1}{2}(\omega^b)^T \mathcal{I} \omega^b + \frac{1}{2} m v^T v \quad (1)$$

$$= \frac{1}{2} (V^b)^T \mathcal{M} V^b \quad (2)$$

where we obtain the second equation by grouping the linear velocity v and body angular velocity ω^b into the body twist V^b , and the *inertia tensor* \mathcal{I} and mass m into the *generalized inertia matrix*:

$$\mathcal{M} = \begin{bmatrix} mI_3 & 0 \\ 0 & \mathcal{I} \end{bmatrix}$$

we can then write our Lagrangian as

$$L = T - V \quad (3)$$

and our equations of motion as

$$\Upsilon = \frac{d}{dt} \frac{\partial L}{\partial \dot{q}} - \frac{\partial L}{\partial q} \quad (4)$$

$$= M(q)\ddot{q} + C(q, \dot{q})\dot{q} + G(q). \quad (5)$$

$M(q)$, the matrix that multiplies the acceleration terms \ddot{q} , is called the *inertia matrix* or the *manipulator inertia matrix* in the context of a robotic manipulator. Note that we have now covered three different uses of the word “inertia” (sorry!):

1. The inertia tensor $\mathcal{I} \in \mathbb{R}^{3 \times 3}$
2. The generalized inertia matrix $\mathcal{M} \in \mathbb{R}^{6 \times 6}$
3. The (manipulator) inertia matrix $M(q) \in \mathbb{R}^{n \times n}$ for n generalized coordinates

We will discuss the inertia tensor further in the next section. For a system composed of multiple links or parts, we have $T = \sum_i T_i$ and $V = \sum_i V_i$.

Problem 1: How does equation (1) simplify if the rigid body is a point mass?

Problem 2: How does equation (1) simplify if motion is limited to a 2D plane?

2 Interpreting the 3D inertia tensor

Consider the simple rigid body shown in figure 1:

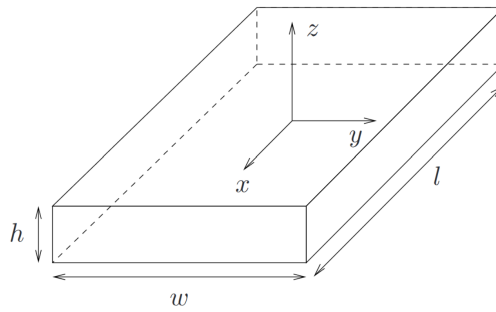


Figure 1: A box with evenly distributed mass.

Let's place the body reference frame at the center of mass of the box, and assume that the box is homogeneous (its mass is distributed evenly). If we rotate the box around some axis ω , we can see that the distribution of mass about the axis will depend on the orientation of ω .

Problem 3: Let $l > w > h$, and consider rotations about the x , y , and z axes. Which rotation should have the highest inertia? Which should have the lowest?

By calculating the velocity of each point in the box due to a rotation and integrating over the mass density times the velocity at each point, it's possible to calculate that the inertia tensor of this box is given by

$$\mathcal{I} = \begin{bmatrix} \frac{m}{12}(w^2 + h^2) & 0 & 0 \\ 0 & \frac{m}{12}(l^2 + h^2) & 0 \\ 0 & 0 & \frac{m}{12}(l^2 + w^2) \end{bmatrix} \quad (6)$$

Problem 4: You've successfully programmed your robot arm to toss the box shown above straight up in the air with speed $v_z = 5$ and a fancy spin given by $\omega = [0, 4, 3]^T$. What is the Lagrangian for the box?

3 Intro to linear dynamics

When working with Lagrangian dynamics, we introduced the “generalized coordinates,” which describe the motion of a system in terms of a set of n scalar coordinates q_i . In control theory, we call these coordinates the *state* of our system, and the set of all possible values of the coordinates form the *state space*. Along the same line, choosing an appropriate set of coordinates is often referred to as *state space modelling*.

Problem 5: *If we want to model the motion of the box being tossed from the previous section, what state space could we choose? What if we want to model an arbitrary toss in any direction and with any spin?*

The easiest kind of systems to analyze are *linear systems*, which have the form

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) \tag{7}$$

$$y(t) = C(t)x(t) + D(t)u(t) \tag{8}$$

where $x \in \mathbb{R}^n$ is the state vector, $u \in \mathbb{R}^m$ is a set of control inputs, $A \in \mathbb{R}^{n \times n}$ describes how the system state will change in the absence of a control input, and $B \in \mathbb{R}^{n \times m}$ describes how the control input affects the system state. If A , B , C , and D do not change over time, we say that the system is *linear time-invariant* or LTI.

In general, we may not be able to directly measure or observe all of the system states, so the observation $y \in \mathbb{R}^{n_o}$ is some linear function of x and u .

Unfortunately, most interesting systems do not have linear dynamics. However, it turns out that nonlinear dynamics can often be approximated with good accuracy in the local neighborhood of a point by “linearizing” them about this point. This means that analyzing the behavior and control of linear systems is still extremely useful, especially since it’s very difficult to analyze nonlinear systems.