

# Practicing Advanced Feedback Control

In this unit we are going to explore basic feedback control on a simplified, 1-dimensional model of the quadrotor-camera. This system is more complicated than the Simple Slide Camera model because it incorporates a dependency between the velocity and a pitch angle as well as a dependency of the output angle,  $\gamma$ , on the pitch angle,  $\theta$ .

## Pitch Slide Camera

Imagine a camera attached to a cart that is allowed to slide along a 1-dimensional track. The track is elevated some height  $h$  above the ground and the camera is pitched at an angle  $\theta$  relative to vertical. There is some target on the ground that the camera can observe; more specifically the camera can measure the angle  $-\pi/2 < \gamma < \pi/2$  from the centerline of the camera to the target. Since the view angle depends on the pitch of the camera,  $\gamma$  is a function of pitch in the form:

$$\gamma = -\left(\tan^{-1}\left(\frac{x_{des}-x}{h}\right) + \theta\right) = \tan^{-1}\left(\frac{x-x_{des}}{h}\right) - \theta$$

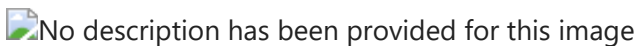
Note that orientation of  $\gamma$  (i.e. the minus sign at the beginning of the expression) is by convention.

The velocity of camera-cart system is also dependent upon the pitch angle; i.e. the larger the pitch, the greater the change in velocity (you could think of this somewhat similar to a segway scooter). The velocity dependence is defined by:

$$\dot{v} = K\sin(\theta)$$

The objective is to issue pitch commands  $\theta_{cmd}$ , in order to move the cart to a position such that  $\gamma \rightarrow \gamma_{des}$  (e.g. if  $\gamma_{des} = 0$ , then the objective is simply to move the cart directly over the target.)

This system is depicted in the below diagram:



This somewhat contrived system can be thought of as a simplified model of the quadrotor and its downward-facing camera. The velocity of a quadrotor is in fact a function of the pitch of the quadrotor, similar to this system. The major simplification here is that the system is constrained to move in 1-dimension, which is not the case for a quadrotor.

Note that the position and velocity of the cart,  $x$  and  $v$ , and position of the target  $x_{des}$  are not directly measured, only  $\gamma$  is measured.

Now we will provide the code necessary to simulate this system

```
In [1]: from __future__ import division, print_function
import numpy as np
import matplotlib.pyplot as plt

_HEIGHT = 1.0
_VEL_CONST = 1.0
_TIME_STEP = 0.1
_THETA_LIMIT = np.pi/4.0
```

## Plant Dynamics, Sensors, and Actuators

the following object contains functions for the plant dynamics, sensing of the target angle  $\gamma$ , and actuator for  $v_{cmd}$

```
In [2]: class PitchSlideCamera():
        '''Object that defines the dynamics of the simple slide-camera'''

        def __init__(self, x_0, v_0, theta_0, x_d, gamma_d=0.0, h=_HEIGHT, k=_VEL_CONST):

            # state variables (hidden)
            self.__x = x_0
            self.__v = v_0

            # reference position (hidden)
            self.__x_d = x_d

            # reference angle (observed)
            self.gamma_d = gamma_d

            # parameters
            self.__h = h
            self.__k = k
            self.__theta_limit = theta_limit

            # control variables (observed, commanded)
            self.__theta = theta_0

        def get_theta(self):
            return self.__theta

        def sense_gamma(self):
            # calculate angle from camera center line to target
            return np.arctan2(self.__x - self.__x_d, self.__h) - self.__theta

        def _get_hidden_position(self):

            return self.__x

        def _get_hidden_position_desired(self):
            return self.__x_d
```

```

def _get_hidden_velocity(self):
    return self.__v

def actuate_theta_command(self, theta_cmd, dt=_TIME_STEP):
    self.__theta = min(self.__theta_limit, max(theta_cmd, -self.__theta_limit))
    self.__v += self.__k*np.sin(self.__theta)*dt
    self.__x += self.__v*dt

```

## Controllers

Functions for control algorithms such as proportional control, proportional-derivative control, etc, as well as any custom controllers you may wish to try

```

In [3]: def p_control(y_err, kp):
        ''' compute the actuator command based on proportional error between output and
        Args:
        y_err: y_des - y where y is the output variable of the plant
        '''

        # TODO: write a proportional control law (hint: it is a single line, very simple)
        cmd = y_err * kp

        return cmd

```

```

In [6]: def pd_control(y_err, y_err_prev, dt, kp, kd):
        '''compute the actuator command based on proportional and derivative error between
        Args:
        y_err: y_des - y where y is the output variable of the plant
        y_err_prev: previous step y_des - y
        '''

        # TODO: write a proportional+derivative control law
        deriv_err = (y_err - y_err_prev)/dt
        cmd = kp * y_err + kd * deriv_err

        return cmd

```

```

In [7]: assert np.isclose(pd_control(0.0, 1.0, 0.1, 1.0, 1.0), -10.0)

```

```

In [8]: def custom_control():
        '''custom-made controller, if you want to develop one
        Args:
        '''
        pass

```

## Simulation Script

below is a script for testing various controllers for the `SimpleSlideCamera` plant as well as plotting the results.

```

In [48]: # Control gains
kp = 0.0008
kd = 0.08

# Control inputs
dt = _TIME_STEP
t_final = 50.0

# initial conditions (position, velocity and target position)
x_0 = 0.0
v_0 = 0.0
theta_0 = 0.0
x_des = 1.0

# create SimpleSlideCamera with initial conditions
pscam = PitchSlideCamera(x_0, v_0, theta_0, x_des)

# initialize data storage
data = dict()
data['t'] = []
data['theta_cmd'] = []
data['theta'] = []
data['err_gamma'] = []
data['x_hidden'] = []
data['v_hidden'] = []
t = 0.0
err_gamma_prev = 0.0
while t < t_final:
    t += dt

    # SENSOR: sense output variable gamma (angle from camera centerline to target)
    err_gamma = pscam.gamma_d - pscam.sense_gamma()

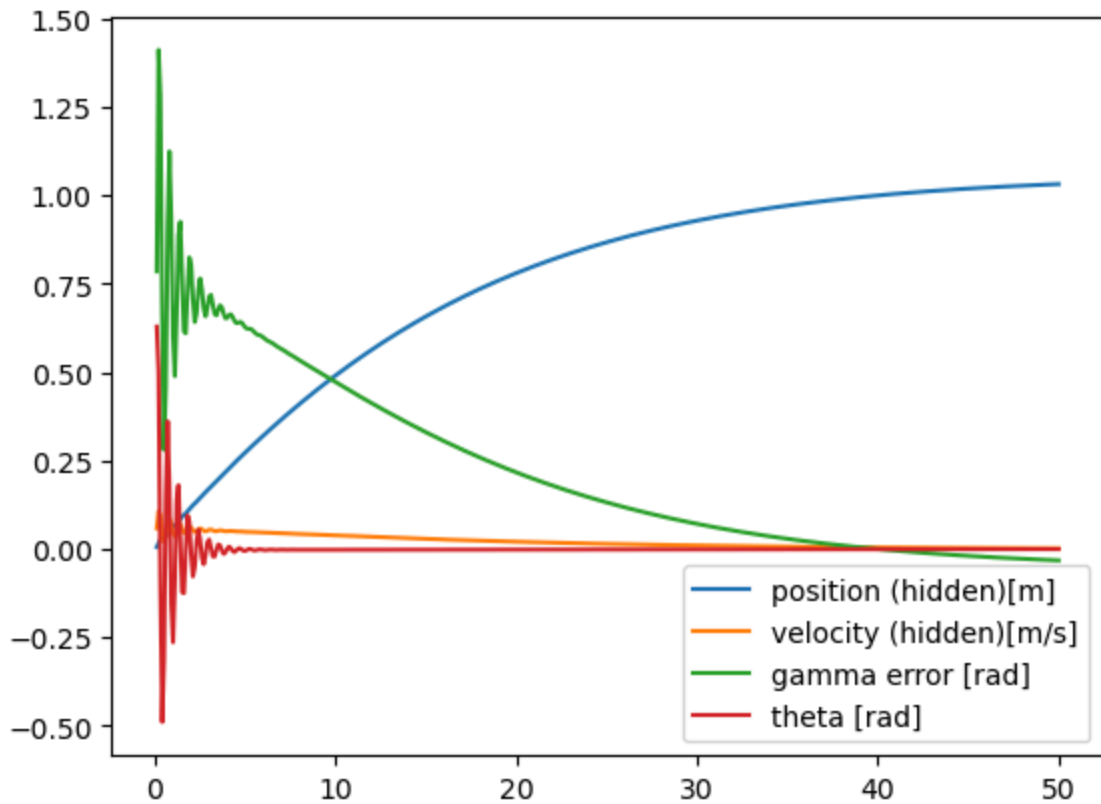
    # CONTROLLER: call theta control algorithm
    theta_cmd = pd_control(err_gamma, err_gamma_prev, dt, kp, kd)

    # ACTUATOR: send velocity command to plant
    pscam.actuate_theta_command(theta_cmd)

    # store data
    err_gamma_prev = err_gamma
    data['t'].append(t)
    data['theta_cmd'].append(theta_cmd)
    data['theta'].append(pscam.get_theta())
    data['err_gamma'].append(err_gamma)
    data['x_hidden'].append(pscam._get_hidden_position())
    data['v_hidden'].append(pscam._get_hidden_velocity())

# Plot Data
handle_position, = plt.plot(data['t'], data['x_hidden'], label='position (hidden)[m]')
handle_velocity, = plt.plot(data['t'], data['v_hidden'], label='velocity (hidden)[m/s]')
handle_err_gamma, = plt.plot(data['t'], data['err_gamma'], label='gamma error [rad]')
handle_theta, = plt.plot(data['t'], data['theta'], label='theta [rad]')
plt.legend(handles=[handle_position, handle_velocity, handle_err_gamma, handle_theta])
plt.show()

```



## Questions

**Q1.** Can you design a controller that is capable of converging the gamma error to zero?

Yes, but it would require a lot of tuning of the parameters. I tried to get the graph above to converge with very small  $k_p$  and  $k_d$  values, but it's still not quite there.

**Q2.** The time scale to make this control converge is on the order of 10s of seconds. Our real drones, which are much more complex control problems, are capable of converging on a target much faster with far less oscillation. Can you give explanations why our drone controllers perform so much better than this controller?

Drones may have more advanced control systems that include integrals (PID), and may in general be able to process things at a much faster speed since they need to make very quick decisions and calculations when in flight.

In [ ]: