

# Dynamic System Modeling and Control Design

## Second Order DT System, Proportional and PD Control

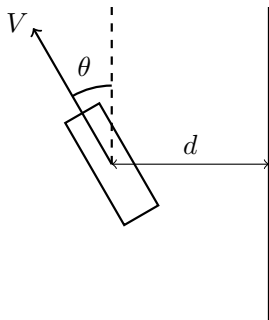
September 18, 2024

# Outline

- 1 Second Order DT System: Line Following Example
- 2 Proportional Controller Shortcomings
- 3 New Controller: Proportional Derivative

## Line Following Example

Consider a line following example illustrated below:



$$d[n] = d[n - 1] + \Delta T V \sin \theta[n - 1],$$

$$\theta[n] = \theta[n - 1] + \Delta T \omega[n - 1],$$

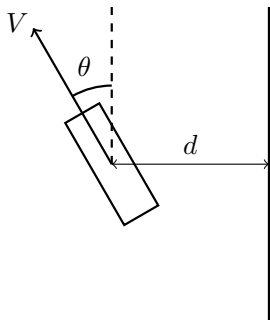
$$\omega[n] = \gamma u[n].$$

Goal: control the angular velocity of the robot to follow the line.

- Assume we have an optical sensor to measure the distance,  $d_m$ .

## Small Adjustments...

Assume that we are operating in a small  $\theta$  regime such that  $\sin \theta \approx \theta$ .



$$d_m[n] = d_m[n-1] + \Delta T V \theta[n-1],$$

$$\theta[n] = \theta[n-1] + \Delta T \omega[n-1],$$

$$\omega[n] = \gamma u[n].$$

Goal: control the angular velocity of the robot to follow the line.

- Assume we have an optical sensor to measure the distance,  $d_m$ .

## Towards a System Equation

Now, we need a system equation with respect to the measured distance. Consider the difference between  $d_m[n]$  and  $d_m[n - 1]$ :

$$d_m[n] = d_m[n - 1] + \Delta TV \theta[n - 1]$$

$$- d_m[n - 1] = d_m[n - 2] + \Delta TV \theta[n - 2]$$

$$d_m[n] - d_m[n - 1] = d_m[n - 1] - d_m[n - 2] + \Delta TV (\theta[n - 1] - \theta[n - 2])$$

With some rearranging:

$$d_m[n] - 2d_m[n - 1] + d_m[n - 2] = \Delta TV (\theta[n - 1] - \theta[n - 2])$$

## Line Following System Equation

Recall that we have:

$$\theta[n] = \theta[n - 1] + \Delta T \omega[n - 1] = \theta[n - 1] + \Delta T \gamma u[n - 1].$$

Therefore,  $\theta[n] - \theta[n - 1] = \Delta T \gamma u[n - 1]$ , and we obtain:

$$d_m[n] - 2d_m[n - 1] + d_m[n - 2] = \Delta T V (\theta[n - 1] - \theta[n - 2])$$

$$\Rightarrow \boxed{d_m[n] - 2d_m[n - 1] + d_m[n - 2] = \Delta T^2 V \gamma u[n - 2]}$$

Now, we need to pick our control signal,  $u[n]$ .

## How About Proportional Control?

First, let's attempt to use proportional control,

$$u[n] = K_p(d_d[n] - d_m[n]).$$

$$\begin{aligned} d_m[n] - 2d_m[n-1] + d_m[n-2] &= \\ &\quad \Delta T^2 V \gamma K_p (d_d[n-2] - d_m[n-2]) \\ \Rightarrow d_m[n] - 2d_m[n-1] + (1 + \Delta T^2 V K_p \gamma) d_m[n-2] &= \\ &\quad \Delta T^2 V \gamma K_p d_d[n-2] \end{aligned}$$

Now, we have a second order difference equation for  $d_m[n]$ .

## Check Yourself: Step Response

Modify the code from last lecture (link) to plot the step response of this second order system:

$$d_m[n] - 2d_m[n-1] + (1 + \Delta T^2 V K_p \gamma) d_m[n-2] = \Delta T^2 V \gamma K_p d_d[n-2],$$

with the following parameters:

- $K_p = 5$ ,
- $\gamma = 1$ ,
- $V = 1$ ,
- $\Delta T = 0.01$ ,
- `simulation_time = 25`,

by defining a new transfer function. Be prepared to show the plot!



## Homogeneous Solution of Second Order DT System

For a second order DT system, the general solution is given by:

$$d_m[n] = C_1\lambda_1^n + C_2\lambda_2^n,$$

where  $\lambda_1, \lambda_2$  are natural frequencies,  $C_1, C_2$  are coefficients determined by the initial conditions.

The homogeneous solution is given by:

$$\lambda^n - 2\lambda^{n-1} + (1 + \Delta T^2 V K_p \gamma)\lambda^{n-2} = 0$$

$$\lambda^2 - 2\lambda + (1 + \Delta T^2 V K_p \gamma) = 0$$

$$\Rightarrow \boxed{\lambda = 1 \pm j\sqrt{\Delta T^2 V K_p \gamma}}$$

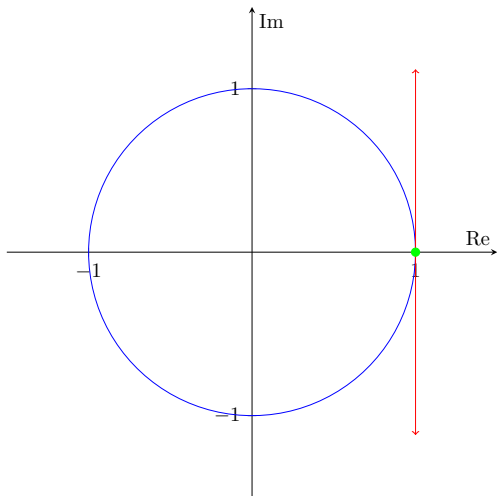
## A Complex Result...

The natural frequencies for the proportional controller are,

$$\lambda = 1 \pm j\sqrt{\Delta T^2 V K_p \gamma}.$$

- $\lambda_1, \lambda_2$  will be complex numbers
- ...with magnitude strictly great than 1!
- This system is always unstable regardless of  $K_p$ .

# Root Locus Plot



$$\lambda = 1 \pm j\sqrt{\Delta T^2 V K_p \gamma}.$$

- When  $K_p = 0$ ,  $\lambda = 1$ .
- As  $K_p \rightarrow \infty$ , moves vertically away from real axis.

# Proportional-Derivative (PD) Controller

New controller: the proportional-derivative (PD) controller:

$$u[n] = K_p (d_d[n] - d_m[n]) + K_d \left( \frac{d_d[n] - d_d[n-1]}{\Delta T} - \frac{d_m[n] - d_m[n-1]}{\Delta T} \right)$$

This controller not only cares about the relative distance, but also the rate of change.

# System Equation with Proportional Derivative Control

From the original system equation,

$$d_m[n] - 2d_m[n - 1] + d_m[n - 2] = \Delta T^2 V \gamma u[n - 2],$$

we can plug in our new PD controller:

$$d_m[n] - 2d_m[n - 1] + d_m[n - 2] = \Delta T^2 V \gamma \left[ K_p (d_d[n - 2] - d_m[n - 3]) + K_d \left( \frac{d_d[n - 2] - d_d[n - 3]}{\Delta T} - \frac{d_m[n - 2] - d_m[n - 3]}{\Delta T} \right) \right]$$

We will skip the algebra necessary to rearrange this equation...

## Third Order Characteristic Equation

We obtain the following characteristic equation:

$$\begin{aligned} d_m[n] - 2d_m[n-1] + (1 + \Delta T^2 V \gamma K_p + K_d V \gamma \Delta T) d_m[n-2] - K_d V \gamma \Delta T d_m[n-3] \\ = (\Delta T^2 \gamma V K_p + K_d \Delta T V \gamma) d_d[n-2] - K_d \Delta T V \gamma d_d[n-3] \end{aligned}$$

This is a third order difference equation, which has the following solution:

$$d_m[n] = C_1 \lambda_1^n + C_2 \lambda_2^n + C_3 \lambda_3^n.$$

Now there are 3 natural frequencies! Each are a function of  $K_p$  and  $K_d$ .

# Third Order System in Python

```
# Define the system parameters
```

```
Kp = 5  
Kd = 1  
gamma = 1  
V = 1  
dt = 0.01
```

```
# Define the transfer function
```

```
num = np.array([0, 0, dt**2*V*gamma*Kp+Kd*V*gamma*dt, -Kd*V*gamma*dt])  
den = np.array([1, -2, 1+dt**2*V*gamma*Kp+Kd*V*gamma*dt, -Kd*V*gamma*dt])
```

```
# Define our third-order system
```

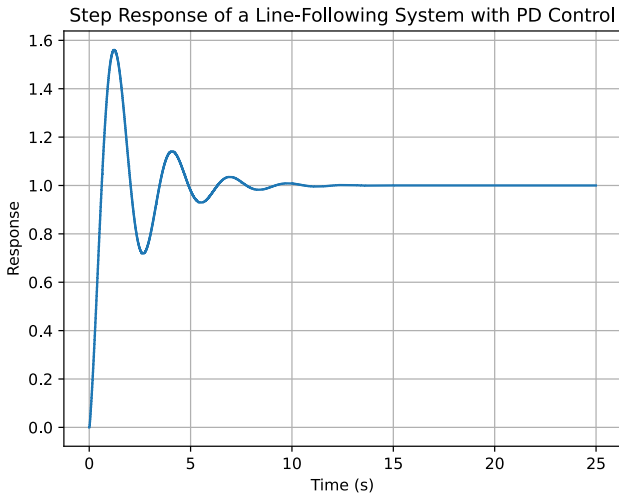
```
system = ctrl.TransferFunction(num,den,dt=dt)
```

```
# Get step response
```

```
time = np.arange(0, 25, dt) # Create the time vector  
_, response = ctrl.step_response(system, T=time)
```

# Third Order System Step Response Plot

$$K_p = 5, K_d = 1, \gamma = 1, V = 1, \Delta T = 0.01$$

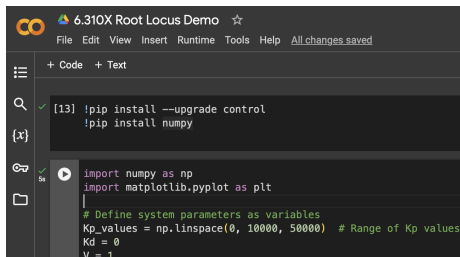




# Generating Locus Plots with Python

Finding closed-form solutions for  $\lambda_1, \lambda_2, \lambda_3$  is possible, but tedious.

Instead, we will generate the locus plots numerically with Python. A Google Colab notebook containing the code to generate the plots in the next few slides is available ([here](#)).



```
6.310X Root Locus Demo ☆
File Edit View Insert Runtime Tools Help All changes saved

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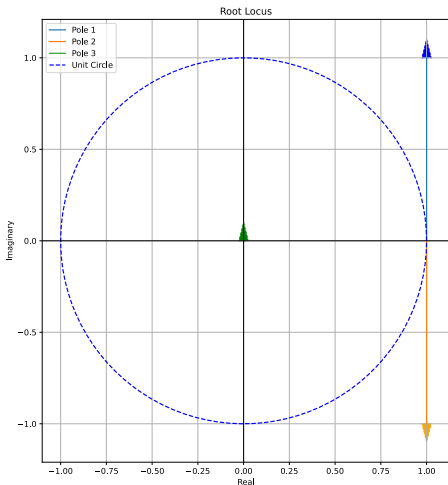
[13] !pip install --upgrade control
!pip install numpy

import numpy as np
import matplotlib.pyplot as plt

# Define system parameters as variables
Kp_values = np.linspace(0, 10000, 50000) # Range of Kp values
Kd = 0
V = 1
```

Case 1: Set  $K_d = 0$ 

As a sanity check, let's generate the root locus plot when  $K_d = 0$ :



## Case 2: Set $K_d = 20$

With a non-zero  $K_d$ , we can see that now there is an optimal  $K_p$  value!

