

# Control Systems - ELEN90055

## *Workshop 4*

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# 1 Introduction

This project aims to create a feedback controller in order to balance an activated inverted pendulum in an upright position whilst rejecting various disturbances.

To achieve this goal, Simulink was used to interface with the robot the pendulum was mounted on, adjusting the controller configuration and parameters, sending commands, and receiving encoder data from the robot's motors as well as data from a gyroscope mounted on the pendulum. This allowed a wide range of variables to be tested to systematically assess their impact on the system behaviour, aiding in design decisions when choosing the final controller configuration.

## 2 System Modelling

### 2.1 Building the Inverted Pendulum Model

The model of the actuated inverted pendulum mounted on the LEGO MINDSTORMS EV3 robot was developed using a combination of physical reasoning from the mathematical model of the system, presenting the form of the model, and experimental data, to identify and approximate suitable model parameters. Our aim was to identify and develop a suitable model to implement feedback control on the actuated inverted pendulum to balance the pendulum in the presence of disturbance impulse disturbance from manually displacing the pendulum and disturbance introduced from the motion of the robot. For this reason, we expect this model to be a simplification of the real system and to inherently differ from the actual system in parameter values. Nonetheless, this model will resemble the actual system in form and will be sufficient to implement feedback control on the inverted pendulum.

### 2.2 Mathematical Model of the Inverted Pendulum

The mathematical basis for the inverted pendulum model is taken from ELEN90055 Problem Set 1. Figure 1 below shows a simplified inverted pendulum sys-

tem:

Figure 1: Inverted Pendulum and DC Motor Systems

In this model, only the ordinary differential equation (ODE) involving the mechanics of the system will be considered. This is because the experimental data obtained was from an unactuated system - only the mechanical dynamics of the system influenced this data. In this system:

- $J = (J_m + ml^2)$ : moment of inertia of the motor and load relative to the motor shaft (kg.m<sup>2</sup>)
- $b$ : viscous damping of shaft (N.s/m)
- $\theta$ : angular displacement of the motor shaft (rad)

Figure 2: Pendulum Model

Using the Newton-Euler equation for the angular acceleration of the pendulum about the motor shaft, we obtain the following equations:

$$\begin{aligned} M &= J\ddot{\theta} \\ mgl \sin(\theta) - T_{damping} &= (J_m + ml^2)\ddot{\theta} \\ mgl \sin(\theta) - b\dot{\theta} &= (J_m + ml^2)\ddot{\theta} \\ mgl \sin(\theta) &= (J_m + ml^2)\ddot{\theta} + b\dot{\theta} \end{aligned}$$

Assuming small values for theta:

$$\Rightarrow mgl \sin(\theta) + b\dot{\theta} = (ml^2)\ddot{\theta} \quad (1)$$

Equation (1) is the ODE governing the free motion of the pendulum after displacement. As can be noted from the form of the expression, the frictional torque due to damping is additive. This choice yields a solution set which most closely resembles the solution set obtained from experimental data in the next section.

### 2.3 Experimental Data vs Calculated Data

Experimental data obtained from displacing the pendulum as shown in Figure 3 where the initial conditions were  $\dot{\theta}(0) = 0$  and  $\theta(0) = \frac{\pi}{2}$  produced the figure below:

Figure 3: Free Pendulum Results

By trial-and-error, the following parameter values for  $m$ ,  $l$ , and  $b$  were chosen to obtain a suitable ODE solution set:

$$\begin{aligned} m &= 0.20 \quad \text{Kg} \\ l &= 0.11 \quad \text{m} \\ b &= 0.004 \quad \text{N.s/m} \end{aligned}$$

where  $g = 9.81 \text{ m/s}^2$ . These values yield the model shown in Figure 4. Figure 4 clearly shows that the pa-

Figure 4: Matching Model to Experimental Results

parameter values selected do not accurately represent the physical system. There is a clear error between the data obtained from the gyro sensor and the data obtained from solving the ODE. However, because the form of calculated model is like the experimental data, we can be confident that developing a controller which meets the provided specifications based on this model will produce a control action suitable for controlling and responding to the dynamics of the physical system. Using the parameter values obtained we can now consider the actuated inverted pendulum system. A relevant subset of the stalled and no-load characteristics of the motor and a small-signal linearized approximation of are shown below:

$$ml^2\ddot{\theta} = \tau - b\dot{\theta} + mgl\theta$$

where

$$\begin{aligned} \tau &= \frac{Kt}{R}(v - K_b\dot{\theta}) \text{ and} \\ \frac{K_t}{R} &= 1.66 \times 10^{-2} \frac{\text{N.m}}{\text{V}} \\ K_b &= 0.297 \frac{\text{V} \cdot \text{sec}}{\text{rad}} \end{aligned}$$

Taking the Laplace Transform of the small-signal linear approximation and accounting for the "percentage full power" units of the motor command (percentage

of 9 V) and gyro sensor and motor encoder angle measurement units (in degrees) yields the following transfer function between the motor command and pendulum angle in degrees:

$$\begin{aligned} G(s) &= \frac{\Theta_{deg}(s)}{V_{\%}(s)} \cdot \frac{\text{degrees}}{\% \text{ full power}} \\ \Rightarrow G(s) &= \frac{9}{100} \cdot \frac{180}{\pi} \cdot \frac{\frac{K_t}{R}}{(s^2 \times ml^2 + s \times (b + \frac{K_t K_b}{R}) - mgl)} \end{aligned}$$

Substituting the system parameters yields the model we used to develop our controller:

$$G(s) = 0.086090 \cdot 0.0242s^2 + 0.008959s - 0.2158$$

### 3 Controller Design

#### 3.1 Performance Requirements

The controller should:

1. Reject impulsive and step disturbances, imparted by manual displacement of the pendulum and turning the robot caused by the drive wheels respectively.
2. Be internally stable.
3. Have a phase margin of  $40^\circ$  or more.
4. Have a complimentary sensitivity function bandwidth of less than 50 rad/sec.

#### 3.2 Controller Implementation

The controller used to achieve these requirements was a phase lead controller, of the form:

$$C(s) = 100 \cdot \frac{(0.1s + 1)}{0.001s + 1}$$

#### Model Considerations to Meet Requirements

- *Meeting the phase requirement.* Observing the phase part of Figure 5, it can be seen that to achieve the required phase margin the controller needs to add phase at crossover.
- *Internal stability.* Figure 6 shows the root locus plot of the plant transfer function detailed in section 2. Because the model contains an unstable

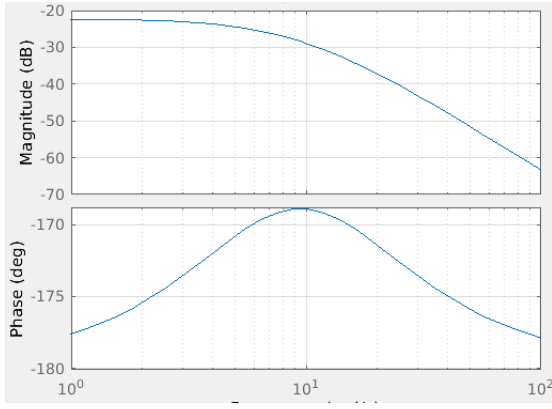


Figure 5: Bode Plot of Plant Model

pole, sufficient gain to stabilise it must be used. Since the relative degree of the model is two, during controller design care was taken to avoid excessive gain values which develop undesirable oscillations in the system.

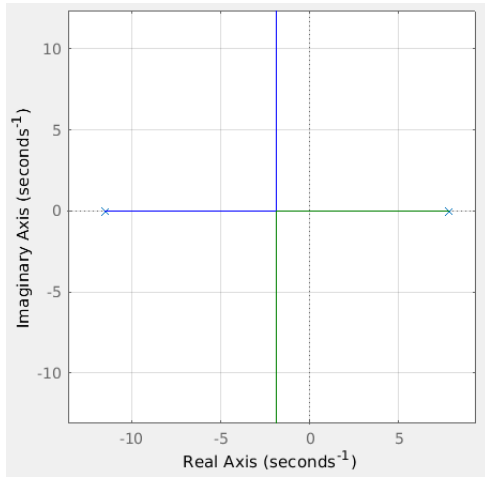


Figure 6: Root Locus of Plant Model

### Loop Shaping

The zero used in the controller was chosen to be placed between the poles in the plant model, this serves to completely stabilise the unstable pole in the plant for sufficiently high  $K$ .

Varying the position of the pole in the nominal controller transfer function  $\frac{0.1s+1}{\tau s+1}$ , Figure 7 shows that phase increases when  $\tau$  is decreased, thus as the pole moves away from the origin. To meet the design requirements, the crossover frequency of the complementary sensitivity function  $T_0$  to be less than 50 rad/sec. Furthermore, to track desired inputs well, the magnitude of the complementary sensitivity function at low

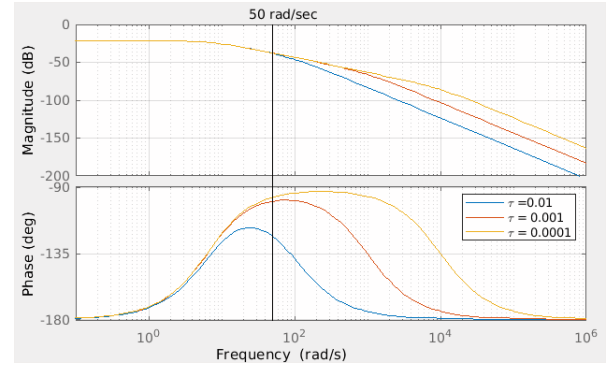


Figure 7: Effect of Pole Placement on Complimentary Sensitivity Function

frequencies should be close to 1 dB. Figure 7 demonstrates the phase margin at maximum crossover frequency (50 rad/sec) for  $\tau = 0.01$  is substantially lower when compared to the smaller values of  $\tau$ .  $\tau = 0.001$  and  $\tau = 0.0001$  have similar phase margins at 50 rad/sec crossover, with even lower discrepancy at lower crossover frequencies. Therefore, since  $\tau = 0.001$  drops off sooner (making it less susceptible to sensor noise), this value was used in the controller.

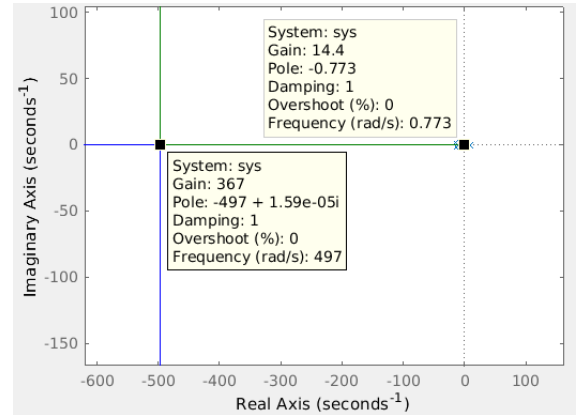


Figure 8: Root Locus of Open Loop System

Finally, based on Figure 8, the system is internally stable and non-oscillatory for  $15 < K < 360$ . In order to meet the sensitivity and physical behaviour requirements,  $K = 100$  was chosen, the resulting bode plot for sensitivity can be seen in Figure ??.

The crossover frequency of the sensitivity function is at approximately 30 rad/sec, which meets the sensitivity bandwidth requirement of 50hz. Furthermore, the magnitude of the sensitivity function for  $\omega < 30$  rad/sec is approximately 1. This is desirable for physical performance, because overly high or

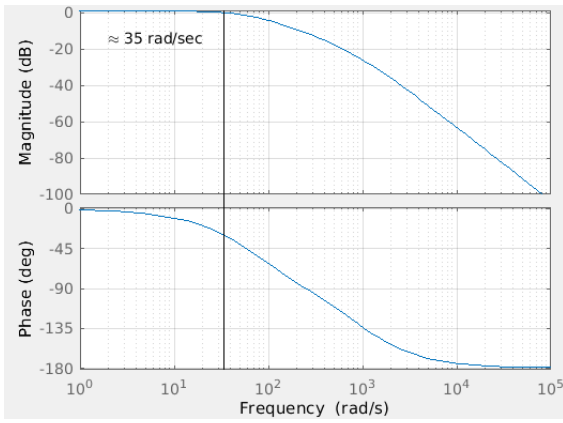


Figure 9: Bode Diagram – Complimentary Sensitivity Function

low sensitivity to perturbations we which to respond to (in the lower range of frequencies) would result in over/undershooting.

**Time Domain Response** The final step in controller design prior to implementation and testing was simulating impulse and step responses of the closed loop system. Figures ?? and ?? show the simulated responses; the behaviour is as anticipated based on prior analysis, showing no oscillation as designed.

## 4 Implementation and testing

### 4.1 Presentation of performance

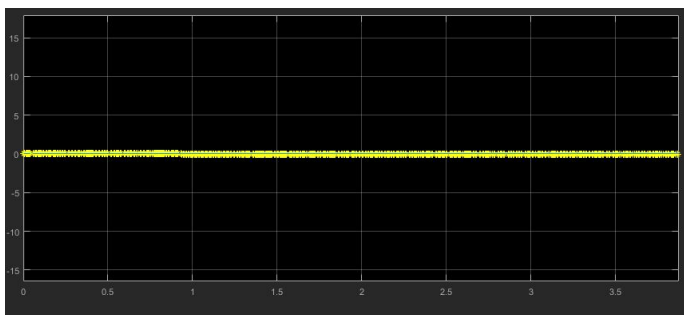


Figure 10: Stable system without disturbance

From figure 9 it can be seen that under optimal conditions the system is perfectly stable. However, as it is possible to balance the arm in this position without the motor being driven this is not a difficult state to maintain. In figure 10 the state the arm returns to when removed from this base state is shown. While the oscillation was not expected the arm is able to maintain

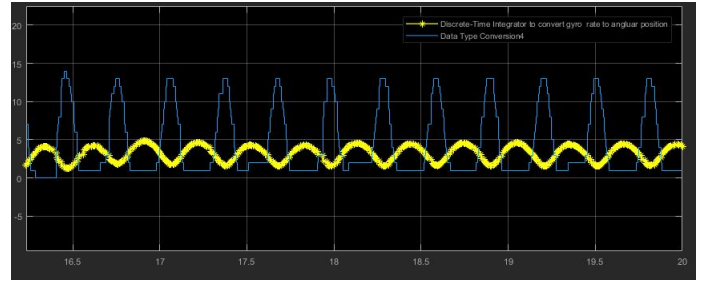


Figure 11: Return state after disturbance

this state indefinitely and can therefore be considered to have remained upright.

### 4.2 Assessment of design criteria

#### 4.2.1 Regulation and Disturbance Rejection

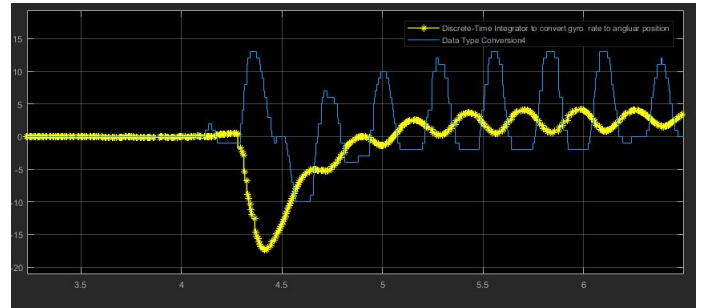


Figure 12: Impulse response

Figure 11 demonstrates the systems ability to recover from an impulse disturbance. As previously mentioned it does not recover from the oscillation, although it is able to maintain the upright position this criterion was considered fulfilled.

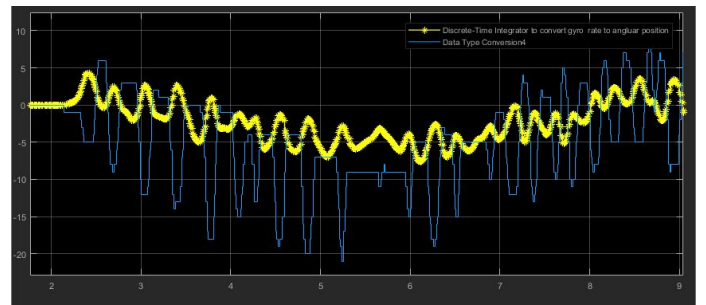


Figure 13: Steering response

For the case of the system in steering mode in figure 12 the result is not so clear. While the high frequency oscillation did not present a problem there also a much

lower frequency oscillation that, when compared with the steering angle in figure 13, can be seen to be related.

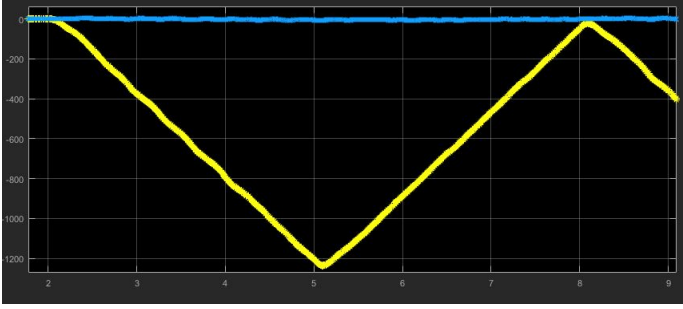


Figure 14: Steering wheel angle

While it may appear that were the turning to continue the system may fail, it can be seen this is not the case in figure 14. Ignoring the high frequency noise, a lower frequency oscillation can be seen opposing the movement of the arm. From this it can be inferred that were the steering to continue then the system would remain an equilibrium where the centripetal force is balanced by the motor.

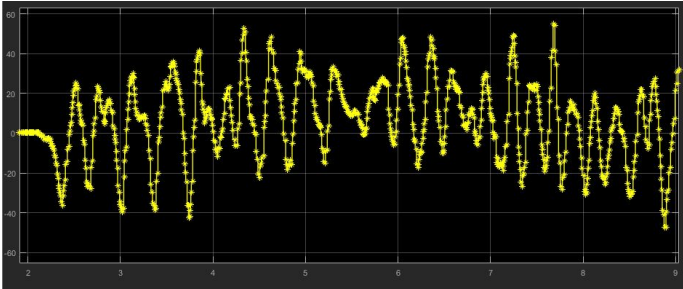


Figure 15: Motor Driving Voltage

#### 4.2.2 Phase Margin

Unfortunately, the phase margin could not be shown to be  $40^\circ$ . Adding a delay in the system corresponding to  $40^\circ$  at the critical frequency was the method used to test this, however, this resulted in the motor voltage repeatedly saturating and very erratic behaviour. This was due to the fact the maximum force that could be delivered by the motor was not taken into account during the design stage, and as such some boundary cases that may have appeared stable are not reliably so in practice.

#### 4.2.3 Bandwidth

The bandwidth requirement could also not be shown to be fulfilled, however, this was due to an inability to test it rather than because it necessarily wasn't. The method of testing this would have been to introduce an  $50\text{rad/s}$  signal and observe that its magnitude was reduced by more than  $\sqrt{2}$ . The issue was that the background oscillation of the system either overwhelmed the response for low disturbance magnitudes, or that the motors force limitation would cause the entire system to fail at high magnitudes. As such this criterium could not be verified one way or the other.

### 4.3 Comparison against simulated system

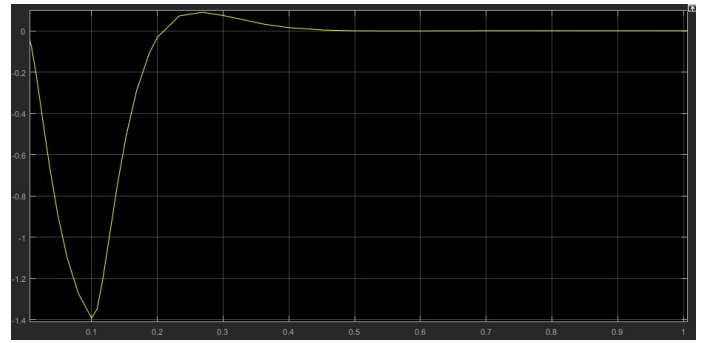


Figure 16: Simulated impulse response

Figure 15 shows the predicted impulse response of the system, and when compared to figure 11 several differences can be observed. Most immediately the oscillation present in the system is absent from the model. In addition, it is clear the system is significantly more damped than the model as on returning to steady state the system displays no overshoot and is significantly slower to do so.

There are several reasons this might be the case. The damping effect is likely primarily due to the unmodelled limitations of the motor rather than any real damping. The motor is unable to apply the full expected force that is applied in the model and this results in a damping like effect when the controller voltage saturates.

The oscillation on the other hand is likely caused by the assumptions made during the modelling. Namely the assumption that the motor is completely fixed. In

reality it could be seen that the motor was able to shift in the robot, and in addition the entire robot would shift with the pendulum. This added motion was not accounted for and likely the primary reason for the oscillation although there were doubtless other contributing factors.

## 5 Conclusion

While we were unable to demonstrate that all of the criteria discussed in 3 were met, we were able to demonstrate through experiment and calculation that the system is robust and stable. Furthermore we analysed and discussed the shortcomings of our control system, and why the realised system behaves differently to the modelled one.

Improvements could certainly be made to the model in order to better fulfil the design criteria. However better modelling the system would exponentially increase the mathematical complexity of the problem, and as such further experimentation may be the best way to find an improved model.