

# Crane Damping

**Applied Robotics - FRTF20**

**HT-2020 Group10**

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# Crane Damping Project

- ☐ **System Overview**
- ☐ **Model Description**
- ☐ **Control Design Approach**
- ☐ **Angle Estimation using IMU**
- ☐ **Network Connection**

# System Overview

System consists of a robot arm that is driven by:

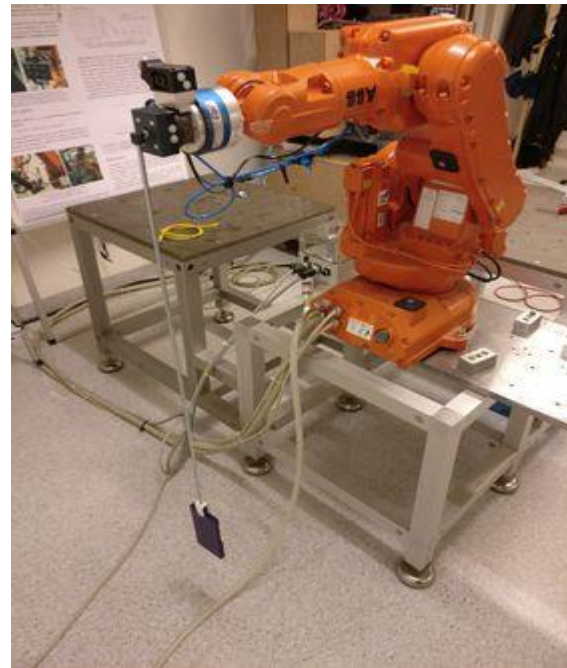
- An **ABB IRB140** robot with 6 joints/links
- A free swinging **pendulum** attached to the pivot point

**IRC5**: Embedded main-computer + PID controller

**Dc motors** at the 6 joints

**Mobile phone** tracker added to the system

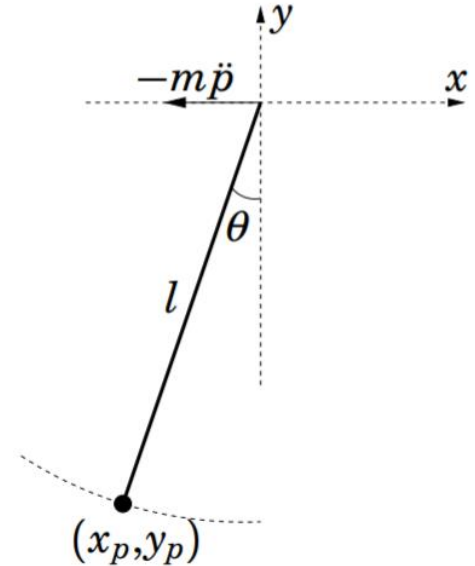
**IMU app** called IMU+GPS-Stream to gather information



# Project Objective

## Damping the load swinging in a planar orbit:

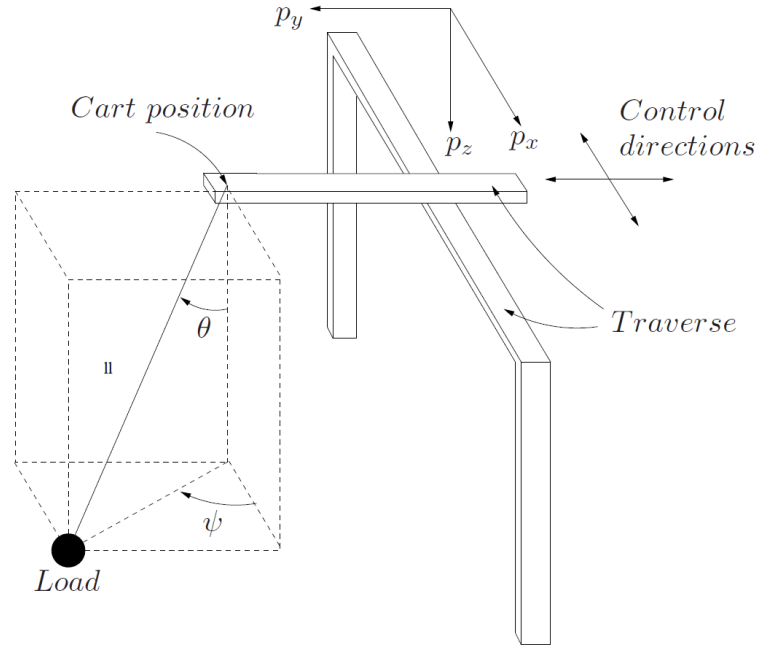
- The main task of our project is to control ABB140 robot arm such that it will be able to damp the oscillations of the swinging pendulum
- using the ABB controller **IRC5** and the LTH-made sensor software **interface ExtCtrl**
- Alternative: Circular swinging process



# Modeling of the Crane Process

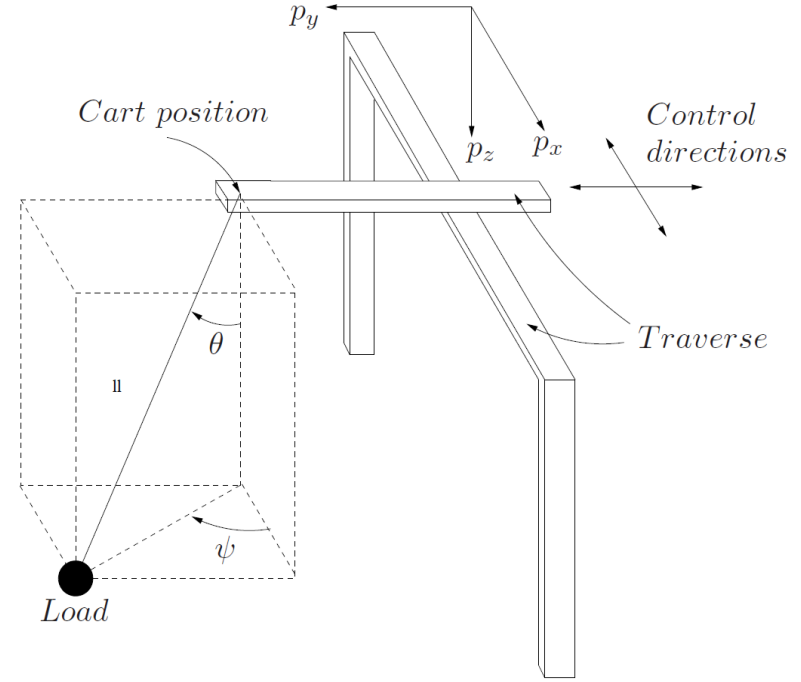
## The Cart-pendulum model

- Our process can be modelled as a **Pendulum** mounted to a **motorized cart**.
- The **cart motion** tries to keep the leaning angle,  $\theta$ , at a small level and damp the swinging pendulum.
- The model should be close to the **real process behaviour**.



# Characteristics of The Physical System

- Pendulum is **oscillatory** without control.
- Dynamics of the pendulum system is **nonlinear**.
- ✓ **Pendulum oscillates back and forth and can be damped by a controller, i.e.** The acceleration of the DC motors tries to keep it at the downward vertical position at **zero degree** by acting on the tilt-angle  $\theta$



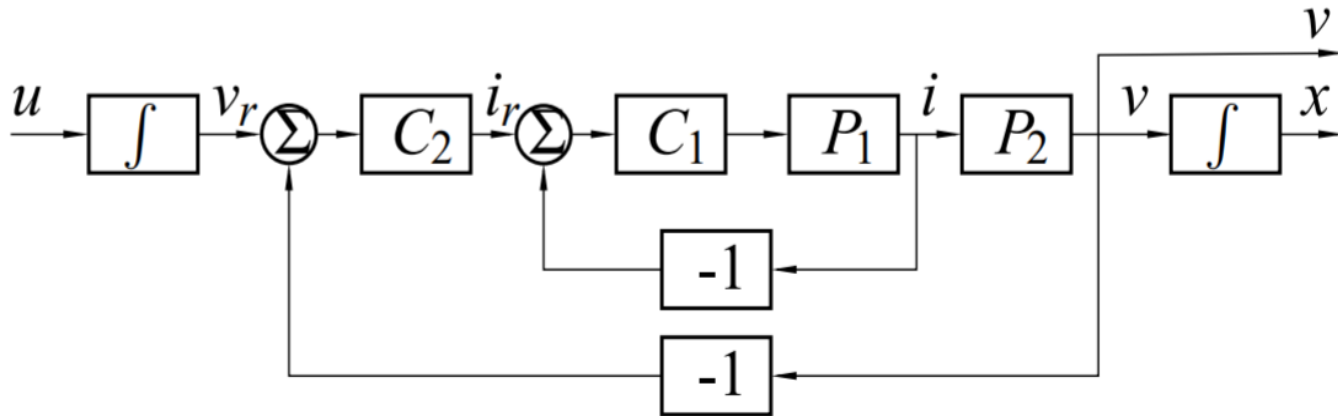
# Modelling the Cart / Cascaded structure

DC-motor is controlled in a cascaded structure

Double integrator

Cart dynamics can be modeled as a **double integrator** from acceleration reference  $u$  to the position  $p$  of the robot arm: **Inner loop** & **Outer loop**

$$\ddot{p} = a_{ref} = u$$



# Kinetic and Potential Energy - Pendulum System

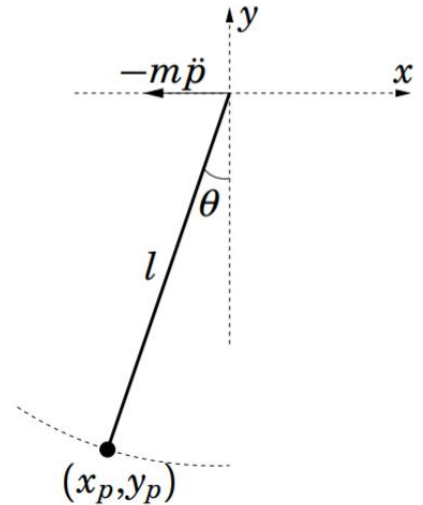
Kinetic and potential energy in the generalized coordinate

$$K = \frac{1}{2}ml^2\dot{\theta}^2$$

$$P = -mgl \cos(\theta)$$

**Lagrangian**

$$L(\theta, \dot{\theta}) = K - P$$



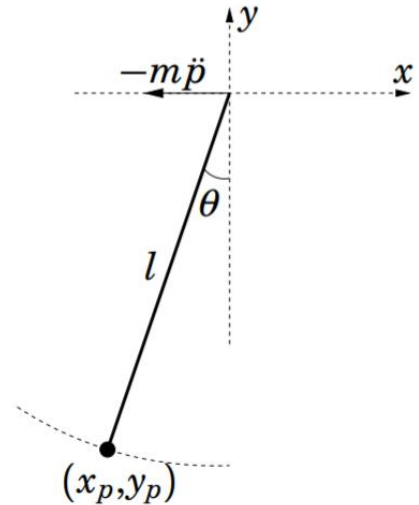


# Dynamic Equations of Pendulum System

Euler-Lagrange 's Law gives the dynamic equation for the pendulum

$$\frac{d}{dt}\left(\frac{\partial L}{\partial \dot{\theta}}\right) - \frac{\partial L}{\partial \theta} = \tau = F_x \frac{\partial r_x(\theta)}{\partial \theta} = m\ddot{p}l \cos \theta$$

$$\ddot{\theta} = \frac{-g}{l} \sin \theta + \frac{\ddot{p}}{l} \cos \theta$$

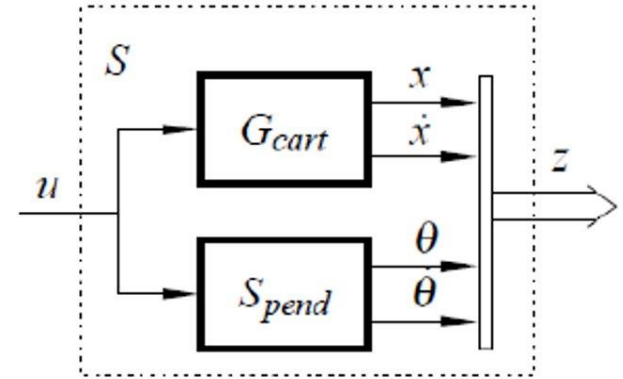


# Resulting Nonlinear Model – Complete System

The two governing equations for the cart-pendulum system

$$\ddot{p} = a_{ref} = u$$

$$\ddot{\theta} = \frac{-g}{l} \sin \theta + \frac{\ddot{p}}{l} \cos \theta$$



# Linearized Pendulum equation of motion

**Linearization:** The model can be linearized by substituting the approximations in the downward position, around  $\theta = 0$

$$\cos \theta \approx 1$$

$$\sin \theta \approx \theta$$

$$\ddot{\theta} = -\frac{g}{l}\theta + \frac{u}{l}$$

**Validation:** This assumption can be reasonably valid if we design the controller such that the pendulum do not deviate too much from the vertically downward equilibrium position.

# The Full State Space System

Linearization of the system dynamics, results in the following state space model

$$z = (z_1, z_2, z_3, z_4)^T = (p, \dot{p}, \theta, \dot{\theta})^T$$

$$\dot{z} = Az + Bu = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & -g/l & 0 \end{pmatrix} z + \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1/l \end{bmatrix} u$$

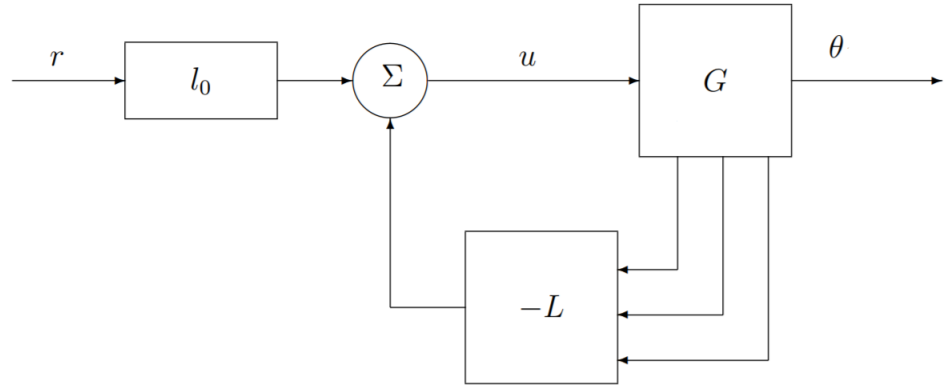
$$y = \begin{pmatrix} p \\ \theta \end{pmatrix} = Cz + Du = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} z + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u$$

# Linear State Feedback Design – LQ Control

## Optimal Feedback Law $u = -Lx$

- Our control design is based on **the linear model**
- LQR design:** computing the optimal gain vector  $L$  that minimizes the cost function  $J$  given the weights  $Q$  and  $R$ :

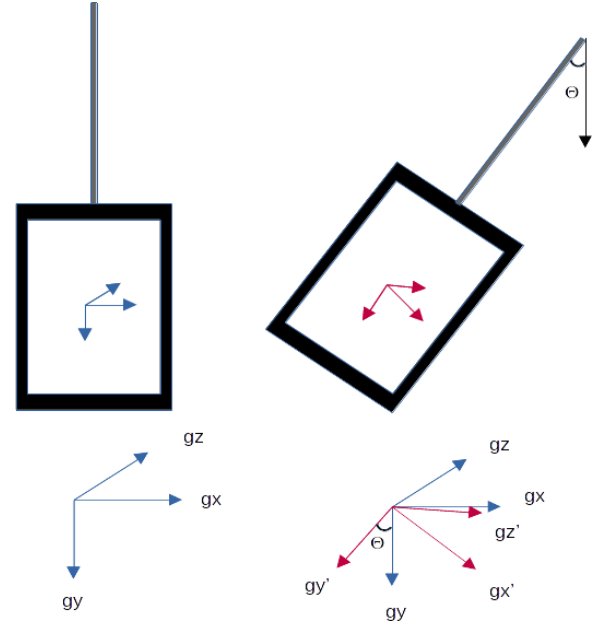
$$J = \int_0^{\infty} (x^T(t)Qx(t) + u^T(t)Ru(t))dt = \int_0^{\infty} (x^T(t)Qx(t) + u^2(t))dt$$



# Angle Estimation using IMU

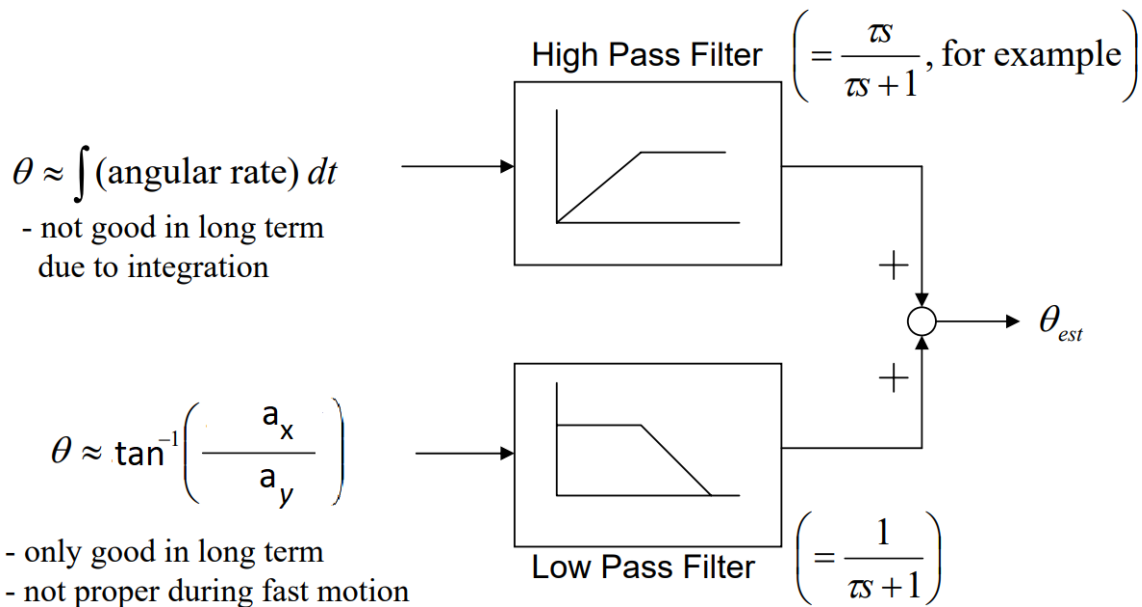
By extracting information from both the **accelerometers** and the **gyro**

**Sensor fusion:** Combining these using **low-pass** and **high-pass** filters, we will create a better estimate of the angle of the pendulum, compared to a naive approach of using only the gyro, or only the accelerometer.



# Angle Estimation IMU - Complimentary Filter

Better estimate by using sensor fusion with Complementary Filter



# Simulink, Python, C and Orca/EXTCTRL

## Network communication to the robot control system

- The model and the optimization problem are described in Matlab/Simulink
- Problem solved by a third-party LTH-made Simulink **ExtCtrl**
- Python and C is the interface that let us communicate to **ExtCtrl** and plot the result

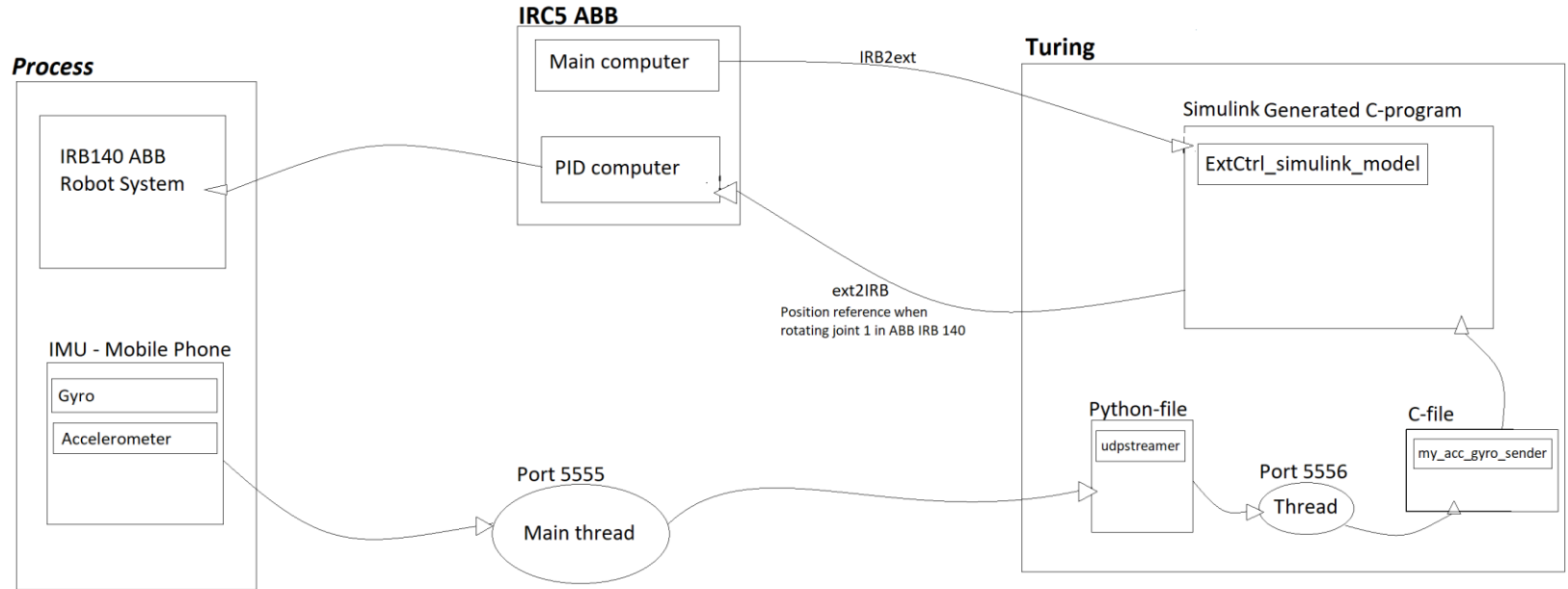


# Communication setup / Network Connection

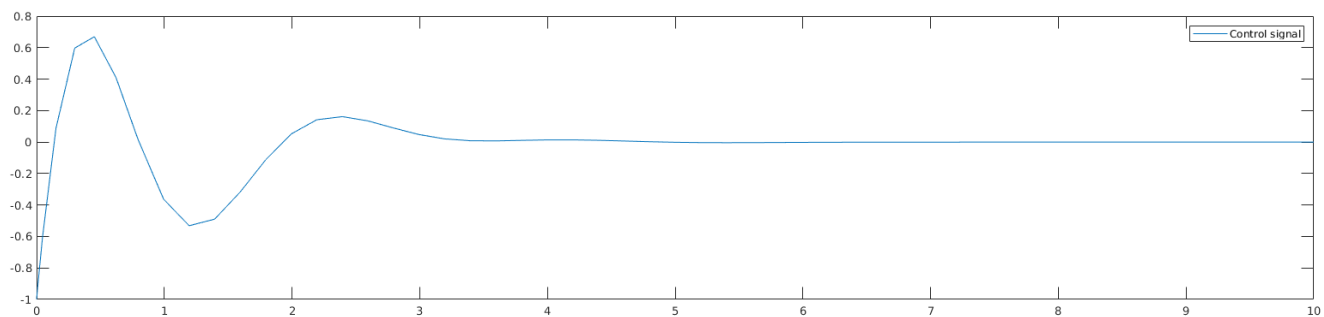
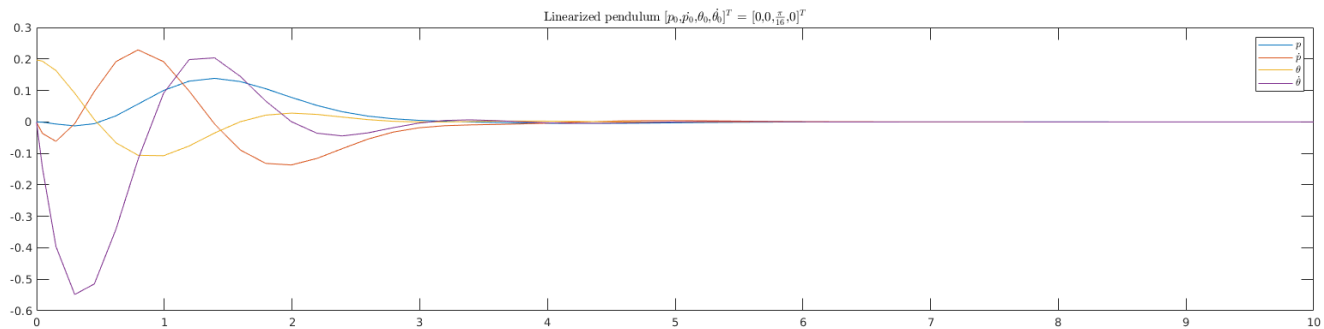
**Three parts** necessary in order to set up a proper communication to the robot:

- **\*.lc-file** to define the variables to send and receive on network
- **Matlab/Simulink model** sets up the connection between Orca and **ExtCtrl**
- **\*.c-file** to manipulate signals or introduce measurement devices

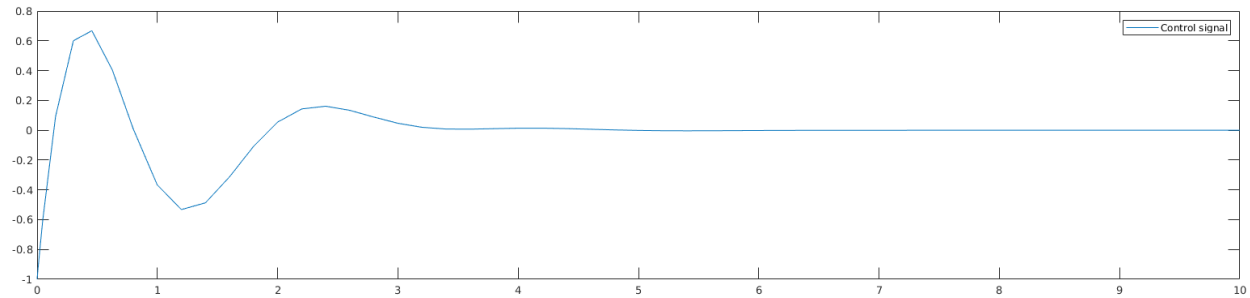
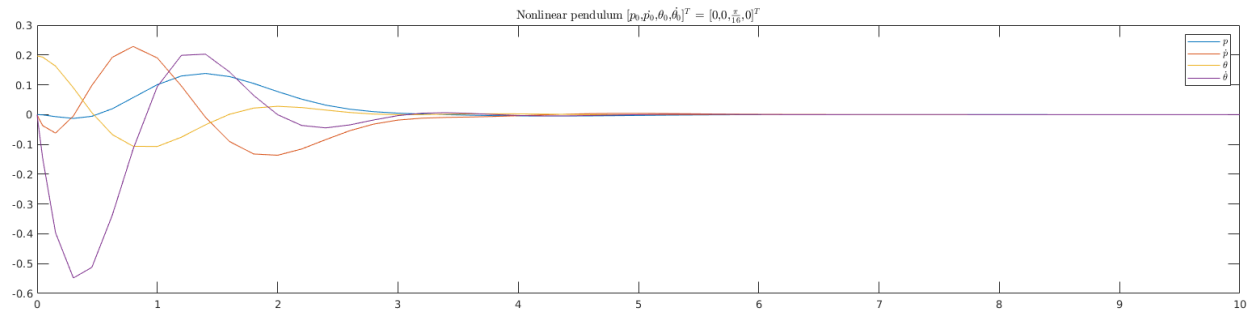
# Communication setup with the robot



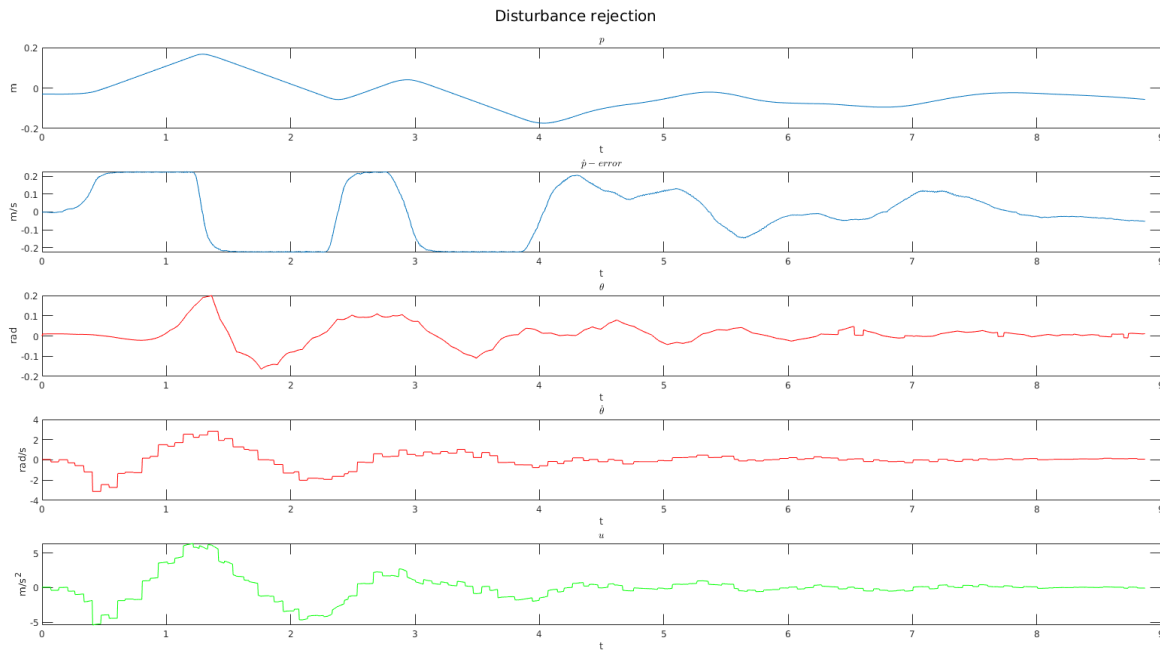
# Linear simulation



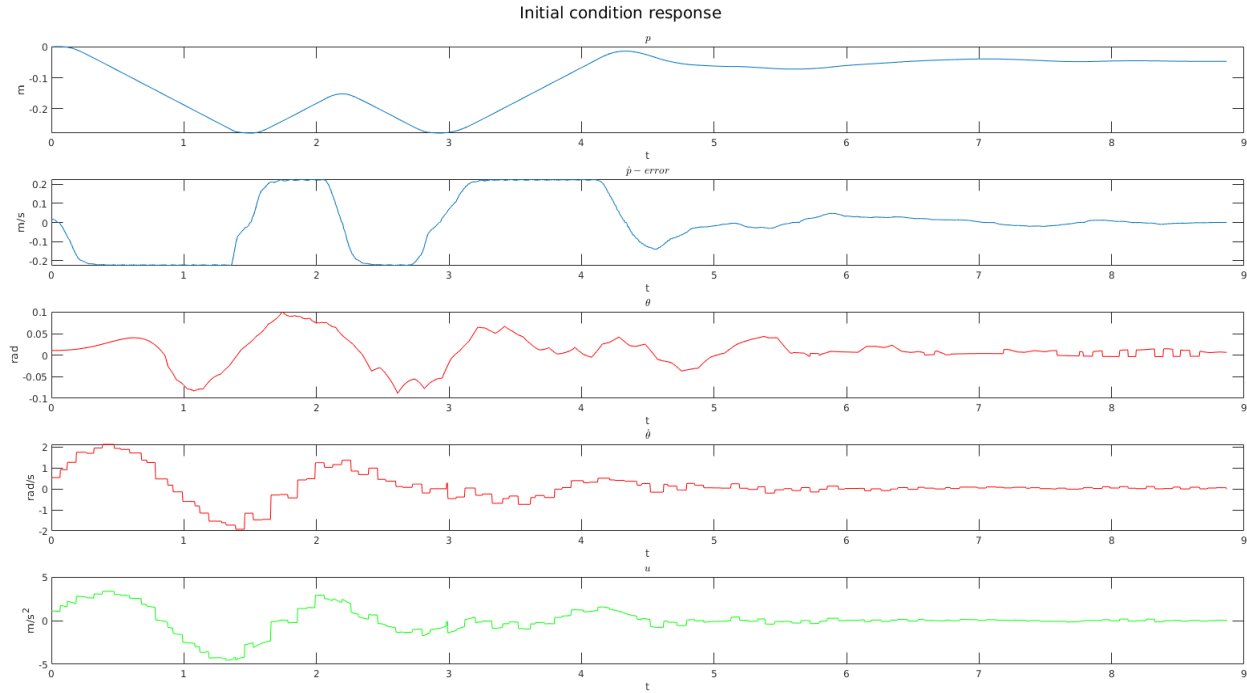
# Nonlinear simulation



# Disturbance Rejection



# Initial value response



# Questions?

**Crane Damping – Group 10**