

Introduction to Modern Controls

Modeling of Dynamic Systems

Why modeling?

Modeling of physical systems:

- a vital component of modern engineering
- often consists of complex coupled differential equations
- only when we have good understanding of a system can we optimally control it:
 - ▶ can simulate and predict actual system response, and
 - ▶ design model-based controllers

Two general approaches of modeling

- based on physics:
 - ▶ using fundamental engineering principles such as Newton's laws, energy conservation, etc
- based on measurement data:
 - ▶ using input-output response of the system
 - ▶ a field itself known as system identification

Part 1: Success Stories for Control

Download Compilation of Success Stories (application/pdf) 

Auto-tuners for PID Controllers (application/pdf) 

Control in Mobile Phones (application/pdf) 

Nonlinear Multivariable Flight Control (application/pdf) 

Advanced Control of Pharmaceutical Crystallization (application/pdf) 

Autopilot for Small Unmanned Aerial Vehicles (application/pdf) 

Controlling Energy Capture from Wind (application/pdf) 

Mobile-Robot-Enabled Smart Warehouses (application/pdf) 

Trip Optimizer for Railroads (application/pdf) 

Control in Stroke Rehabilitation (application/pdf) 

Advanced Control for the Cement Industry (application/pdf) 

Dynamics and Control for Deep-Sea Marine Risers (application/pdf) 

Active Safety Control for Automobiles (application/pdf) 

Advanced Energy Solutions for Power Plants (application/pdf) 

Optimal Ship-Unloading Solutions (application/pdf) 

Automotive Engine-Based Traction Control (application/pdf) 

Automated Manual Transmissions (application/pdf) 

Ethylene Plantwide Control and Optimization (application/pdf) 

Digital Printing Control: Print Shop in a Box (application/pdf) 

Verification of Control System Software (application/pdf) 

Coordinated Ramp Metering for Freeways (application/pdf) 

Advanced Zinc Coating Control in Galvanizing Lines (application/pdf) 

Digital Fly-by-Wire Technology (application/pdf) 

H-infinity Control for European Telecommunication Satellites (application/pdf) 

Road Grade Estimation for Advanced Driver Assistance Systems (application/pdf) 

Controller Performance Monitoring (application/pdf) 

Control for Formula One! (application/pdf) 

Advanced Control Design for Automotive Powertrains (application/pdf) 

Control of NASA's Space Launch System (application/pdf) 

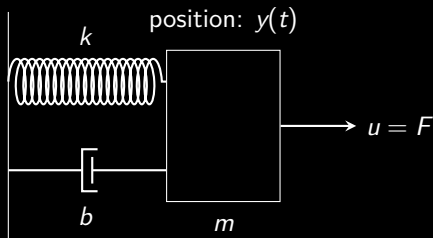
Robust Adaptive Control for the Joint Direct Attack Munition (application/pdf) 

Improved Audio Reproduction with Control Theory (application/pdf) 



- often the tools are combined in practice

Example: Mass spring damper

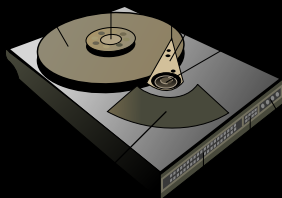


- Newton's second law gives

$$m\ddot{y}(t) + b\dot{y}(t) + ky(t) = u(t), \quad y(0) = y_0, \quad \dot{y}(0) = \dot{y}_0$$

- modeled as a second-order ODE with input $u(t)$ and output $y(t)$

Example: HDD



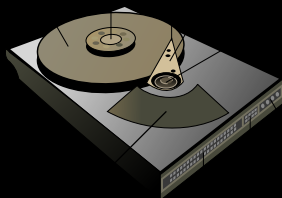
- Newton's second law for rotation

$$\underbrace{\sum_i \tau_i}_{\text{net torque}} = \underbrace{J}_{\text{moment of inertia}} \underbrace{\alpha}_{\text{angular acceleration}}$$

- letting $\theta := \text{output}$ and $\tau := \text{input}$ yields

$$\ddot{\theta} = \alpha = \frac{1}{J} \tau$$

Example: HDD



$$\ddot{\theta} = \alpha = \frac{1}{J}\tau \Leftrightarrow \Theta(s) = \frac{1}{Js^2}T(s)$$

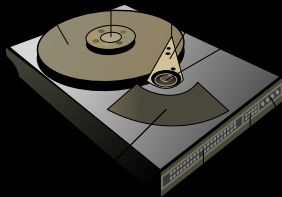
- with damping:

$$\ddot{\theta} + 2\zeta\omega_n\dot{\theta} + \omega_n^2\theta = \kappa\tau \Leftrightarrow \Theta(s) = \frac{\kappa}{s^2 + 2\zeta\omega_ns + \omega_n^2}T(s)$$

- with multiple modes:

$$\ddot{\theta}_i + 2\zeta_i\omega_i\dot{\theta}_i + \omega_i^2\theta_i = \kappa_i\tau \Leftrightarrow \Theta_i(s) = \frac{\kappa_i}{s^2 + 2\zeta_i\omega_is + \omega_i^2}T(s)$$

Example: HDD



$$\ddot{\theta}_i + 2\zeta_i\omega_i\dot{\theta}_i + \omega_i^2\theta_i = \kappa_i\tau \Leftrightarrow \Theta_i(s) = \frac{\kappa_i}{s^2 + 2\zeta_i\omega_i s + \omega_i^2}T(s)$$

- final model:

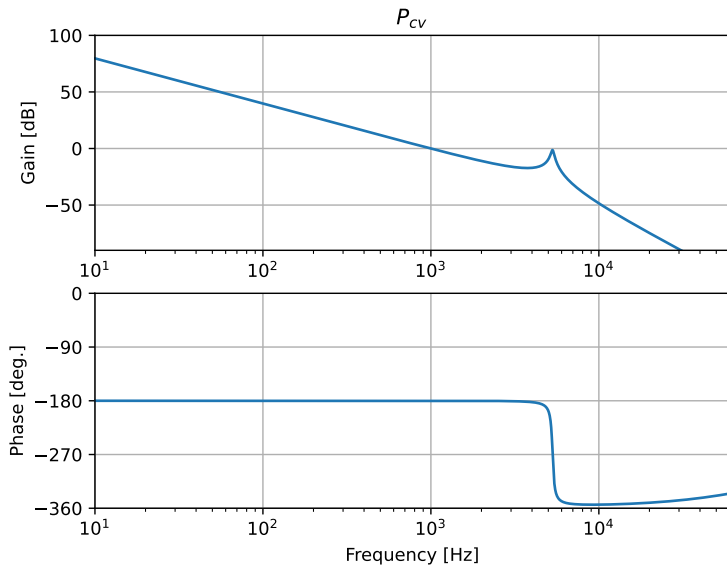
$$\Theta(s) = \sum_{i=1}^n \frac{\kappa_i}{s^2 + 2\zeta_i\omega_i s + \omega_i^2}T(s)$$

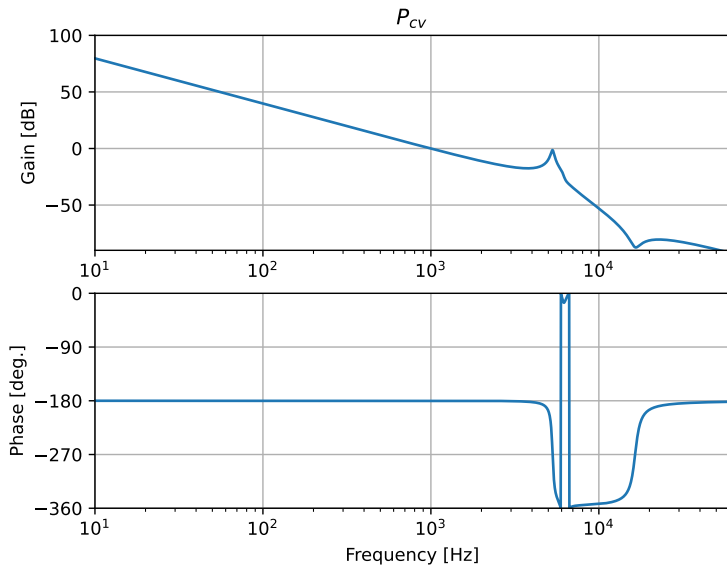

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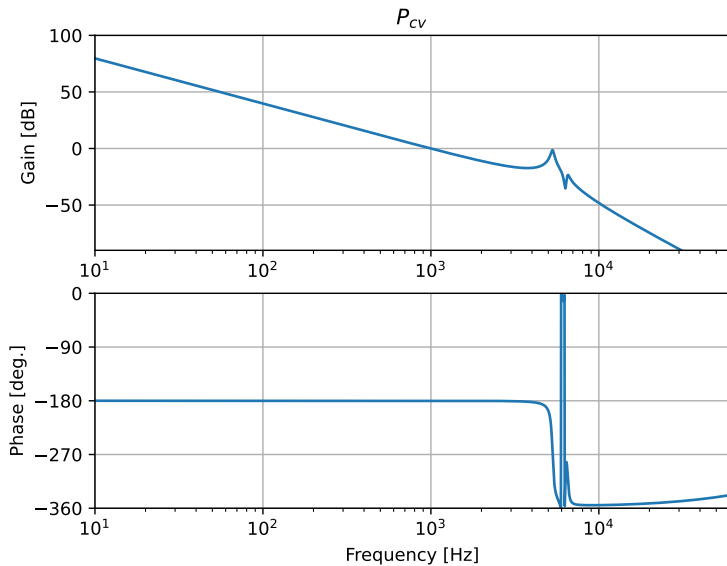
import numpy as np
import matplotlib.pyplot as plt
from scipy import signal
import control as ct

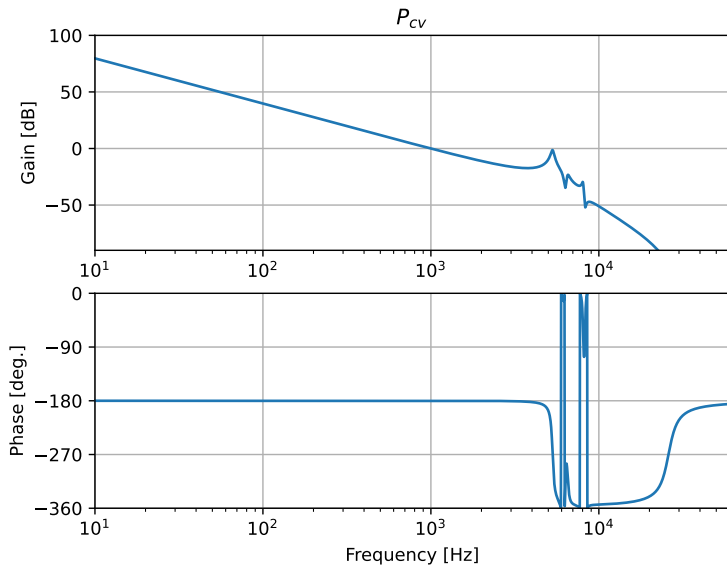
num_sector = 420 # Number of sector
num_rpm = 7200 # Number of RPM
Kp_vcm = 3.7976e+07 # VCM gain
omega_vcm = np.array([0, 5300, 6100, 6500, 8050, 9600, 14800, 17400,
                      21000, 26000, 26600, 29000, 32200, 38300, 43300,
                      ↪ 44800]) * 2 * np.pi
kappa_vcm = np.array([1, -1.0, +0.1, -0.1, 0.04, -0.7, -
                      0.2, -1.0, +3.0, -3.2, 2.1, -1.5, +2.0, -0.2,
                      ↪ +0.3, -0.5])
zeta_vcm = np.array([0, 0.02, 0.04, 0.02, 0.01, 0.03, 0.01,
                     0.02, 0.02, 0.012, 0.007, 0.01, 0.03, 0.01, 0.01,
                     ↪ 0.01])
Sys_Pc_vcm_c1 = ct.TransferFunction([], [1]) # Create an empty
↪ transfer function
for i in range(len(omega_vcm)):
    Sys_Pc_vcm_c1 = Sys_Pc_vcm_c1 + ct.TransferFunction(np.array(
        [0, 0, kappa_vcm[i]]) * Kp_vcm, np.array([1, 2 * zeta_vcm[i] *
        ↪ omega_vcm[i], (omega_vcm[i]) ** 2]))

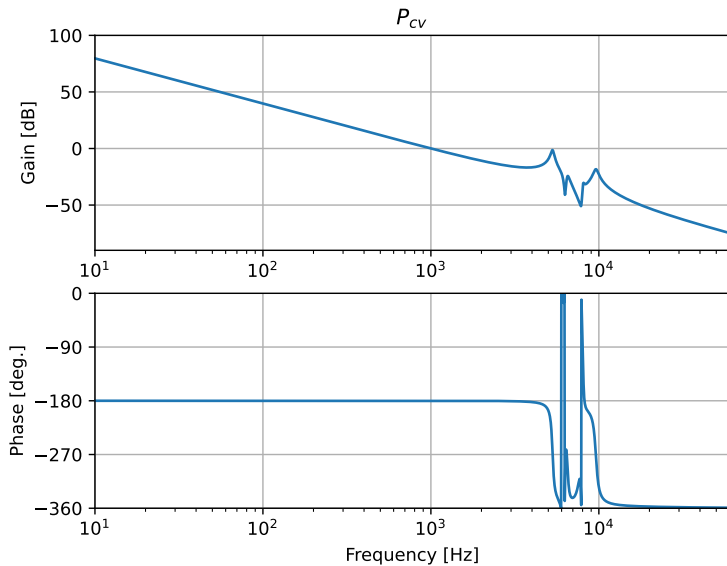
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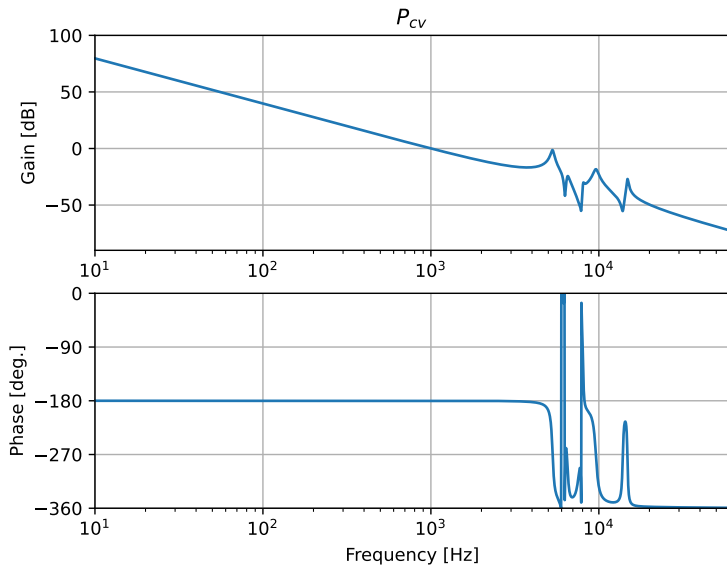



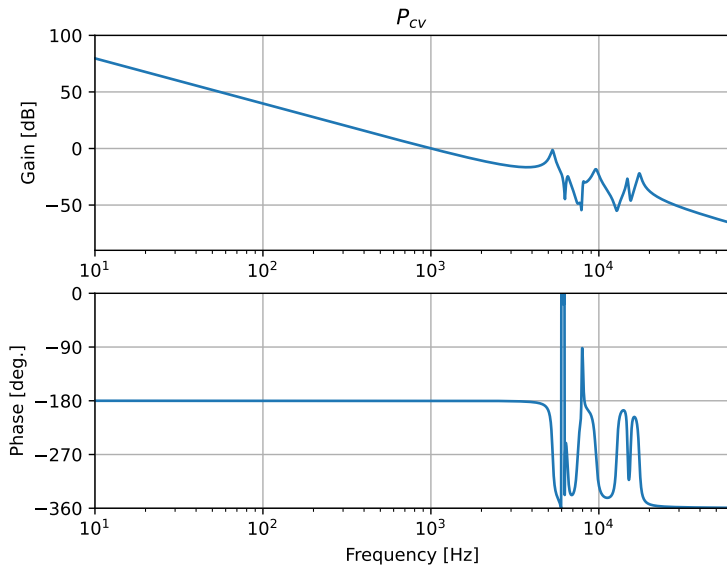


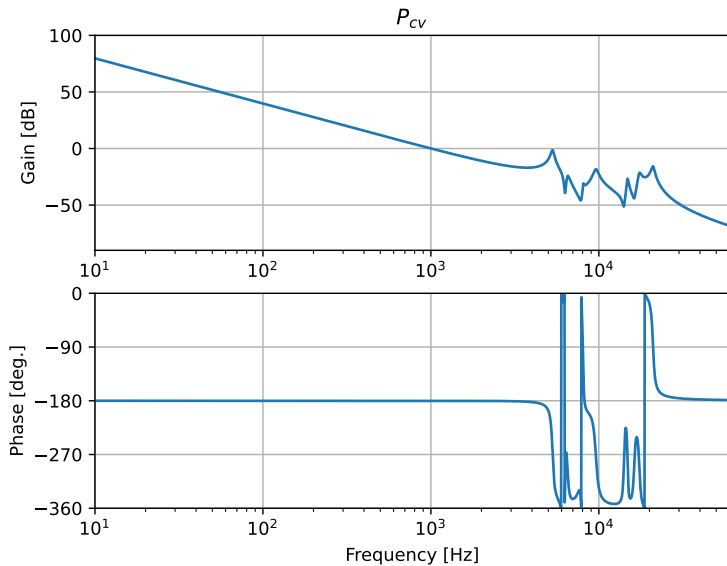


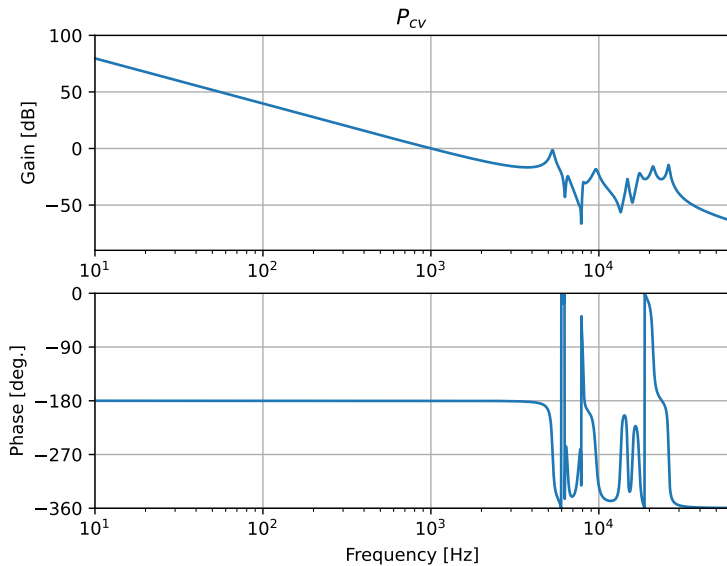


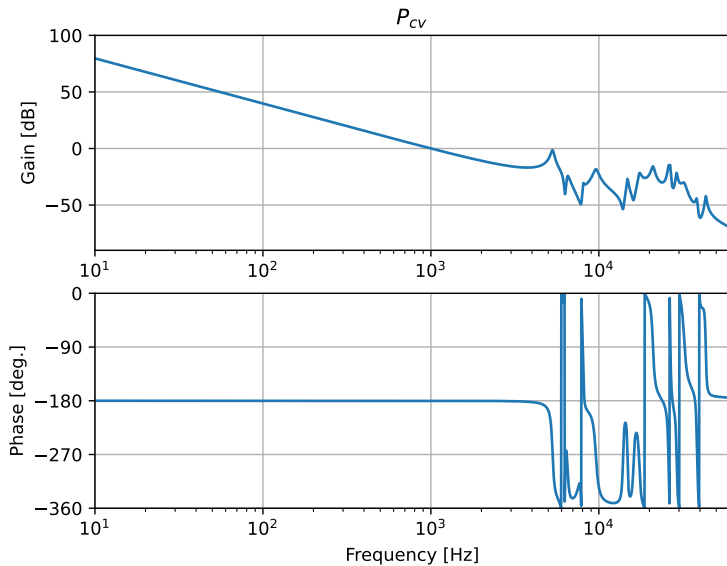






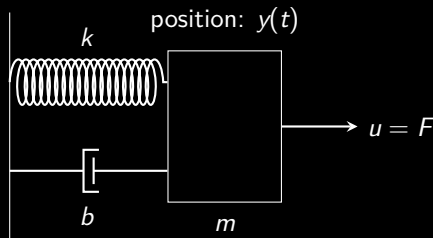




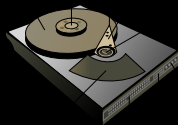


Models of continuous-time systems

- modeled as differential equations:



$$m\ddot{y}(t) + b\dot{y}(t) + ky(t) = u(t), \quad y(0) = y_0, \quad \dot{y}(0) = \dot{y}_0$$



$$\ddot{\theta}_i + 2\zeta_i\omega_i\dot{\theta}_i + \omega_i^2\theta_i = \kappa_i\tau \Leftrightarrow \Theta_i(s) = \frac{\kappa_i}{s^2 + 2\zeta_i\omega_i s + \omega_i^2}T(s)$$

Models of continuous-time systems

General continuous-time systems:

$$\frac{d^n y(t)}{dt^n} + a_{n-1} \frac{d^{n-1} y(t)}{dt^{n-1}} + \cdots + a_0 y(t) = b_m \frac{d^m u(t)}{dt^m} + b_{m-1} \frac{d^{m-1} u(t)}{dt^{m-1}} + \cdots + b_0 u(t)$$

with the initial conditions $y(0) = y_0, \dots, y^{(n)}(0) = y_0^{(n)}$.

Models of discrete-time systems

General discrete-time systems

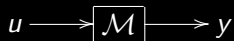
- inputs and outputs defined at discrete time instances $k = 1, 2, \dots$
- described by ordinary difference equations in the form of

$$y(k) + a_{n-1}y(k-1) + \dots + a_0y(k-n) = b_mu(k+m-n) + \dots + b_0u(k-n)$$

Example: bank statements

- $x(k+1) = (1 + \rho)x(k) + u(k), x(0) = x_0$
- k – month counter; ρ – interest rate; $x(k)$ – wealth at the beginning of month k ; $u(k)$ – money saved at the end of month k ; x_0 – initial wealth in account

Model properties: static v.s. dynamic, causal v.s. acausal



Model \mathcal{M} is said to be

- *memoryless* or *static* if $y(t)$ depends only on $u(t)$
- *dynamic* (has memory) if y at time t depends on input values at other times
- e.g.: $y(t) = \mathcal{M}(u(t)) = \gamma u(t)$, $y(t) = \int_0^t u(\tau) d\tau$, $y(k) = \sum_{i=0}^k u(i)$
- *causal* if $y(t)$ depends on $u(\tau)$ for $\tau \leq t$
- *strictly causal* if $y(t)$ depends on $u(\tau)$ for $\tau < t$, e.g.: $y(t) = u(t - 10)$

Linearity and time-invariance

The system \mathcal{M} is called

- *linear* if satisfying the *superposition* property:

$$\mathcal{M}(\alpha_1 u_1(t) + \alpha_2 u_2(t)) = \alpha_1 \mathcal{M}(u_1(t)) + \alpha_2 \mathcal{M}(u_2(t))$$

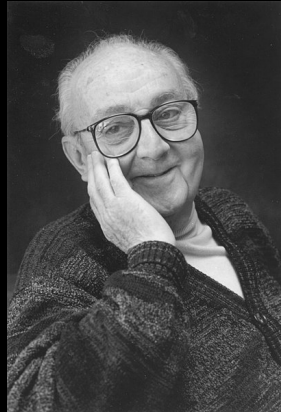
for any input signals $u_1(t)$ and $u_2(t)$, and any real numbers α_1 and α_2

- *time-invariant* if its properties do not change with respect to time
- e.g., $\dot{y}(t) = Ay(t) + Bu(t)$ is linear and time-invariant
- $\dot{y}(t) = 2y(t) - \sin(y(t))u(t)$ is nonlinear, yet time-invariant
- $\dot{y}(t) = 2y(t) - t\sin(y(t))u(t)$ is time-varying
- assuming the same initial conditions, if we shift $u(t)$ by a constant time interval, i.e., consider $\mathcal{M}(u(t + \tau_0))$, then \mathcal{M} is time-invariant if the output $\mathcal{M}(u(t + \tau_0)) = y(t + \tau_0)$

George Box

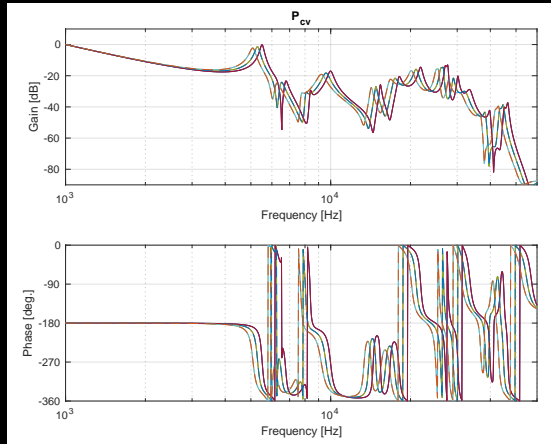
“All Models are Wrong, but Some are Useful”

- statistical models always fall short of the complexities of reality but can still be useful nonetheless
- a dynamic system may simply be too complex (consider the neural system of human brains)
- or there are inevitable hardware uncertainties such as the fatigue of gears or bearings in a car

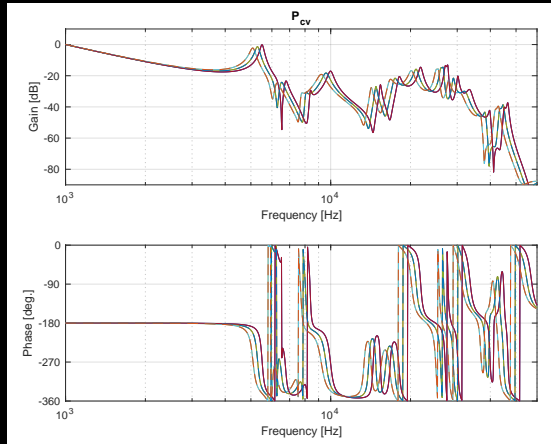
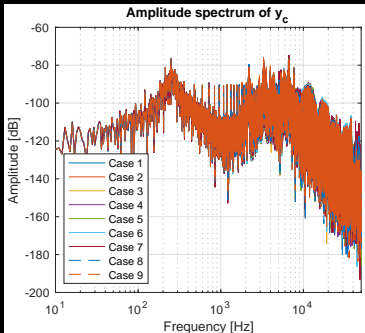


Example (HDDs under perturbation)

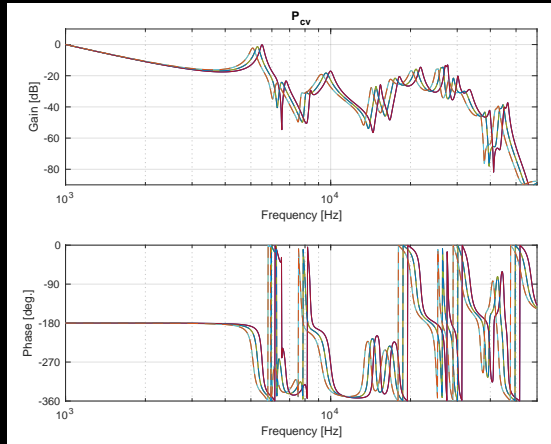
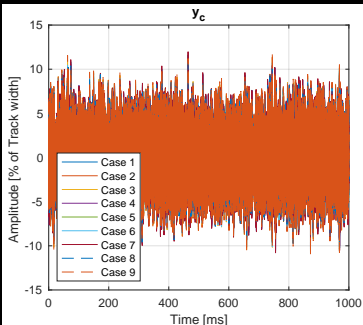
- temperature influence
- manufacturing variations
- but, control works!



Example (HDDs under perturbation)



Example (HDDs under perturbation)



Benchmark Problem for Magnetic-Head Positioning Control System in HDDs

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Abstract: In the “Cloud Era”, the data capacity of the hard disk drives (HDD) must grow to develop the cloud. As a result, we must improve the positioning accuracy of the magnetic head in the HDD. To encourage research about magnetic-head positioning control, we release a benchmark problem that works on MATLAB. This benchmark problem enables us to simulate the magnetic-head positioning control system for the latest HDDs with our designed controller. In this paper, a control design method with the decoupling filter is also presented for this benchmark problem.

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Keywords: Precision control, Data storage, Positioning systems, Actuators, Servos.

1. INTRODUCTION

According to a major data-storage device manufacturer, Western Digital, the future of the cloud service is dependent on the hard disk drive (HDD) capacity growth because demands for the data capacity in the cloud service are rapidly increasing. To solve this issue, we are going to improve the accuracy of a magnetic-head positioning control system so that size of bits for data stored on a disk decreases (Yanagishi and Atsumi (2006), Akhmetshin and Puyang (2015), Atsumi (2016), Nedraoui (2019)).

In order to encourage research about the magnetic-head positioning control, a technical committee consisting of representatives of major universities and an HDD manufacturer with HDD service research in Japan has developed an open-source HDD benchmark problem and released it on the MarkWorks File Exchange (Atsumi et al. (2022)). This benchmark problem enables us to simulate the magnetic-head positioning control system for the latest HDDs used in the cloud storage with our designed controller.

This paper presents the details of the benchmark problem and a control design method with the decoupling filter.

2. HARD DISK DRIVE

Figure 1 shows a picture of the HDD with the cover opened. The HDD consists of a voice coil motor (VCM),

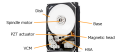


Fig. 1. Hard disk drive.



Fig. 2. Magnetic-head positioning system.



Fig. 3. Feedback control system.

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Example (HDDs under perturbation)

