

# myfbcd.py

## Automated Feedback Controller Design

<https://github.com/shimodatakaki/mypy>

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# 1.1 Overview

## Automated Feedback Controller Design:

Exploiting FRF results, the objective is to find a linear FB controller that satisfies desired linear/quadratic, **convex/concave** constraints **for all given FRFs**, i.e.

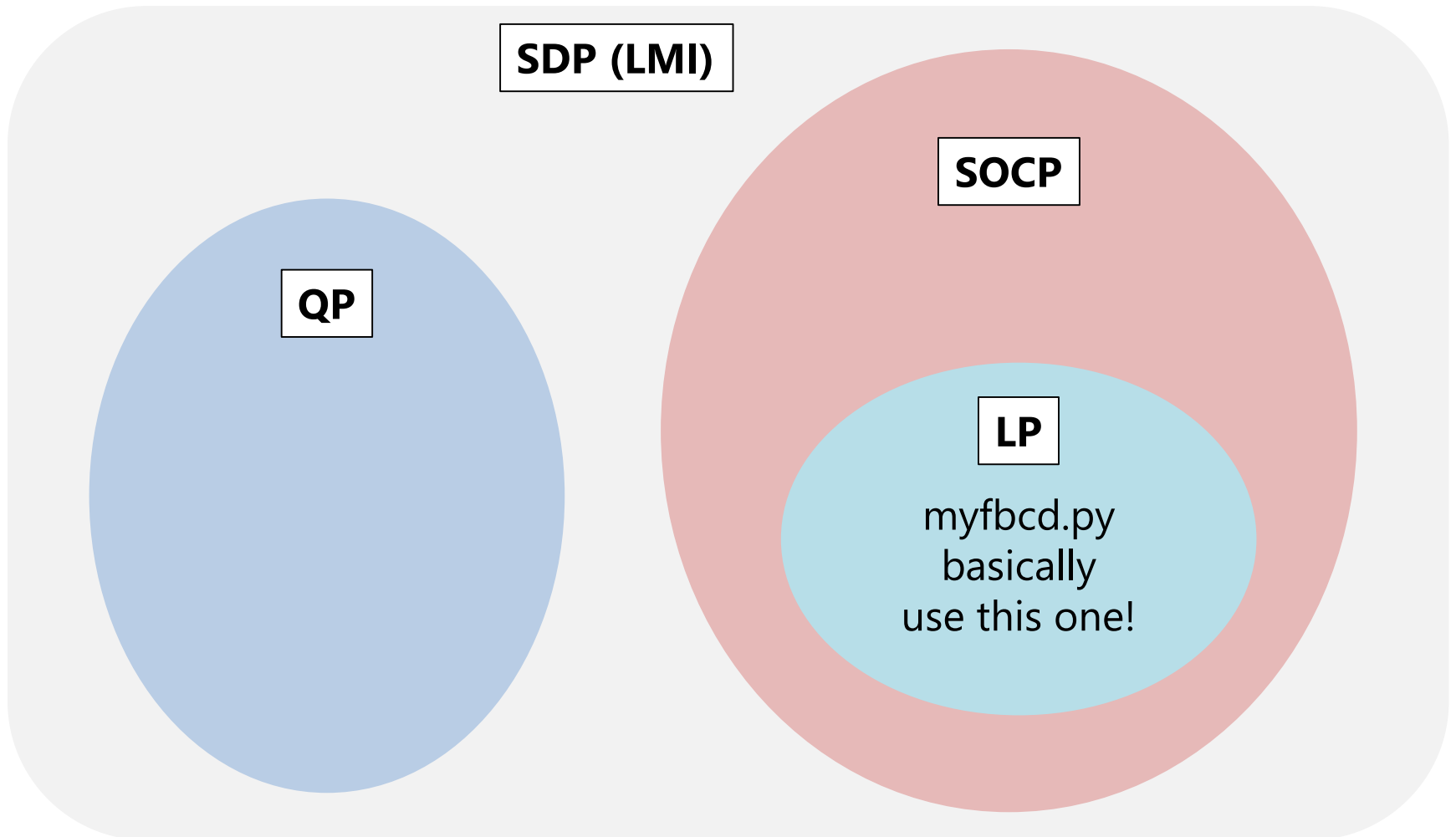
- (1) Gain-Crossover Linear Inequalities,
- (2) Phase Margin Linear Inequalities,
- (3) Gain Margin Linear Inequalities,
- (4) Second Phase Margin Linear Inequalities,
- (5) Gain Minimum/Maximum Linear Inequalities,**
- (6) Stability Margin (Disk) Concave Inequalities via CCCP method,**
- (7) Robust Stability Quadratic Inequalities (using socp or sdp),
- (8) Nominal Performance (Disk) Concave Inequalities via CCCP method.**

NOTE: (5), (6), and (8) are important constraints.

Default Controller: PIDs + 50 FIRs (3+50 variables).

# 1.2 Convex Optimization

Linear/Quadratic Convex Optimization Methods:



## 1.3: Controllers and FRFs

Given FRF  $g$ , the open loop transfer function  $L$  is:

$$L(j\omega_k) = g(j\omega_k)C(j\omega_k)$$

$$C(j\omega_k) = \phi(j\omega_k)^T \rho$$

where  $C$  is the controller,  $\rho$  is the controller gain vector, and  $\Phi$  is the basis transfer function vector.

For example,

$$\phi(s_k) = [1, 1/s_k, s_k/(1 + 0.01s_k)]^T$$

$$\rho = [K_P, K_I, K_D]^T$$

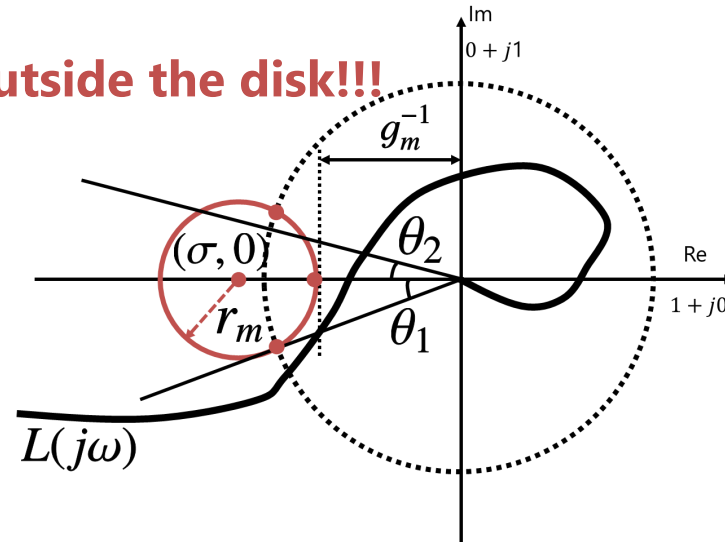
# 1.4 Concave Optimization

(6) and (8) constraints says

"All of Nyquist curves must be outside the disk given for gain-margin, phase-margin, or nominal-sensitivity".

But **out-of-disk constraints are typical concave constraints.**

**Outside the disk!!!**

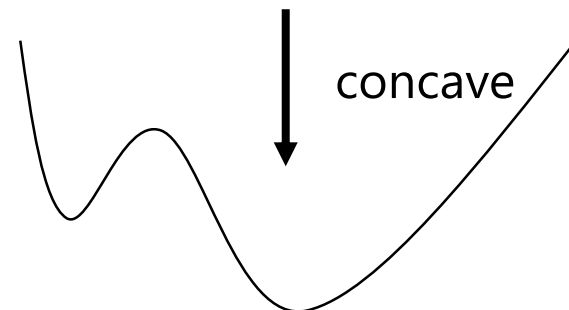
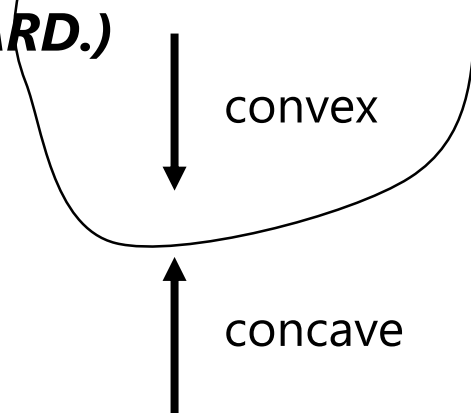


Concave problems cannot be solved by any types of convex solvers in polynomial-time,

i.e. find  $r$  s.t.  $r^2 > 1 \rightarrow$  the result is **primal infeasible.**

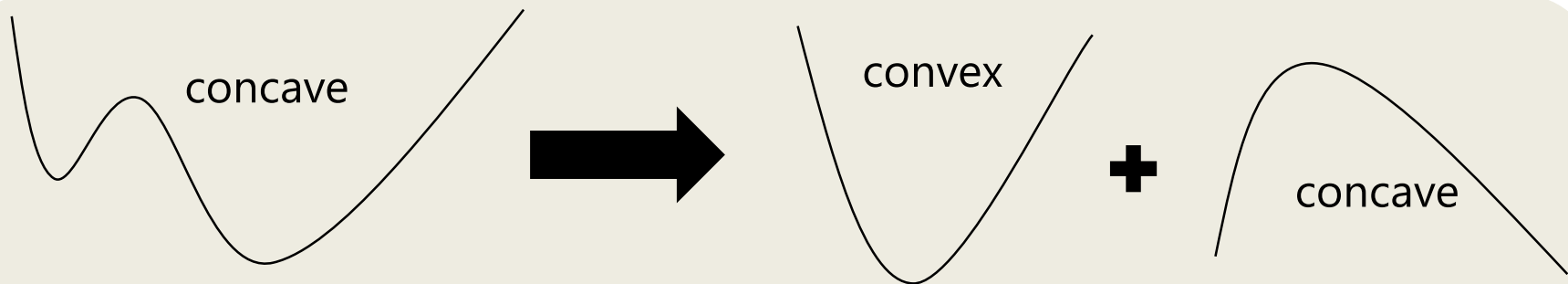
**(Concave optimization is generally**

**NP-HARD.)**



# 1.5 Concave-Convex Procedure (CCCP)

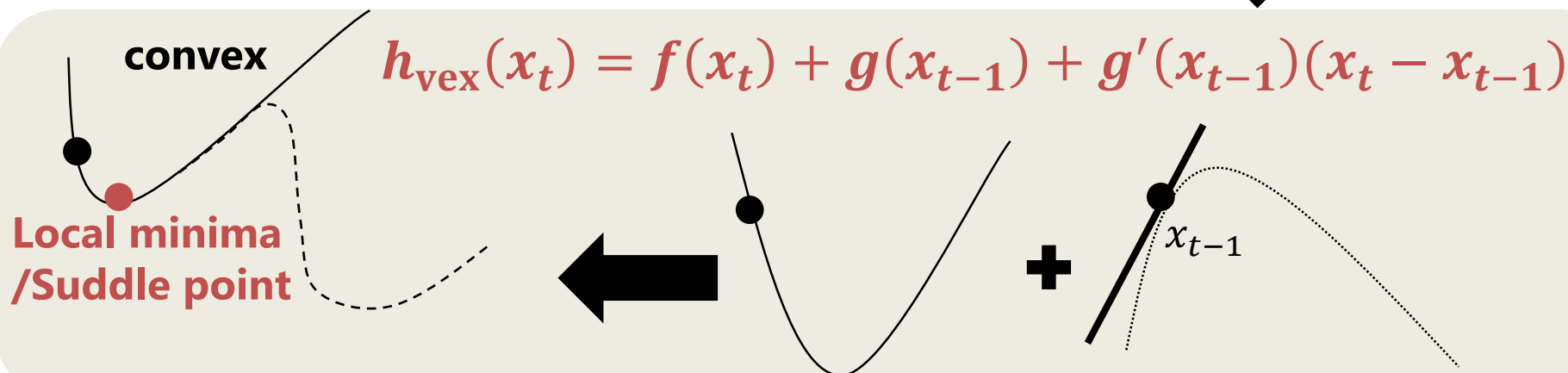
Concave functions can be split into convex and concave functions



$$h_{\text{cave}}(x) = f(x) + g(x)$$

The concave problem is now reduced into a convex problem!

Taylor Series of  $g$

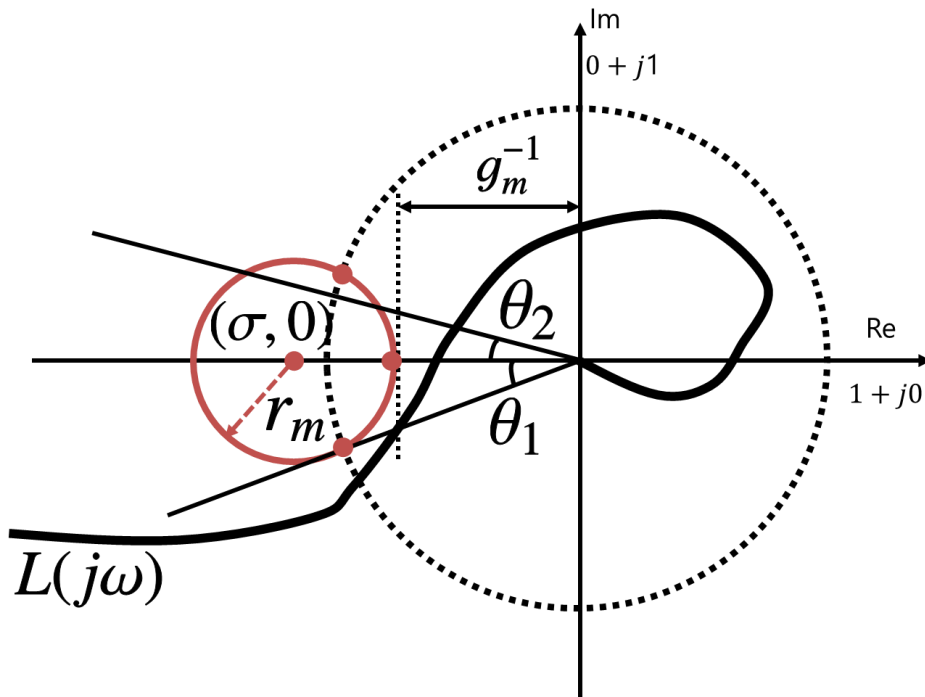


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## 2.1 Stability Margin (Disk) Concave Inequalities via CCCP method



**All of  $L(j\omega_k)$  should be outside the disk.**

$$r_m - |L(j\omega_k) - \sigma| \leq 0$$

for all  $k$

CCCP

like  $x_0x + y_0y \geq r$

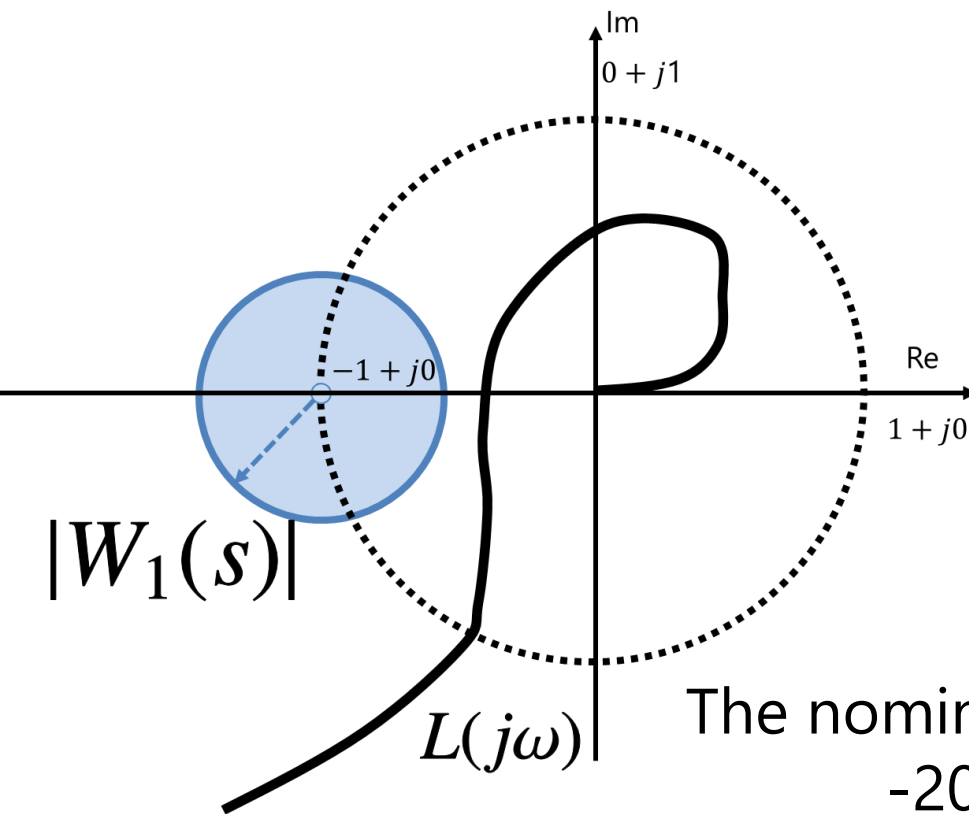
$$\mathbf{G}_k \boldsymbol{\rho} \leq \mathbf{h}_k$$

for all  $k$

$$G_k = -\Re \left[ \frac{L_{t-1}(j\omega_k)^H}{|L_{t-1}(j\omega_k)|} \phi(j\omega_k)^T \right]$$

$$h_k = -r_m - \Re \left[ \frac{L_{t-1}(j\omega_k)^H}{|L_{t-1}(j\omega_k)|} \right] \sigma$$

## 2.2 Nominal Performance (Disk) Concave Inequalities via CCCP method



$$|W_1(j\omega_k)S(j\omega_k)| \leq 1$$

for all  $k$

$$|W_1(j\omega_k)| = \left( \frac{\omega_{ns}}{\omega_k} \right)^m$$

The nominal sensitivity  $\leq 0$  dB at  $\omega_{ns}$ , and

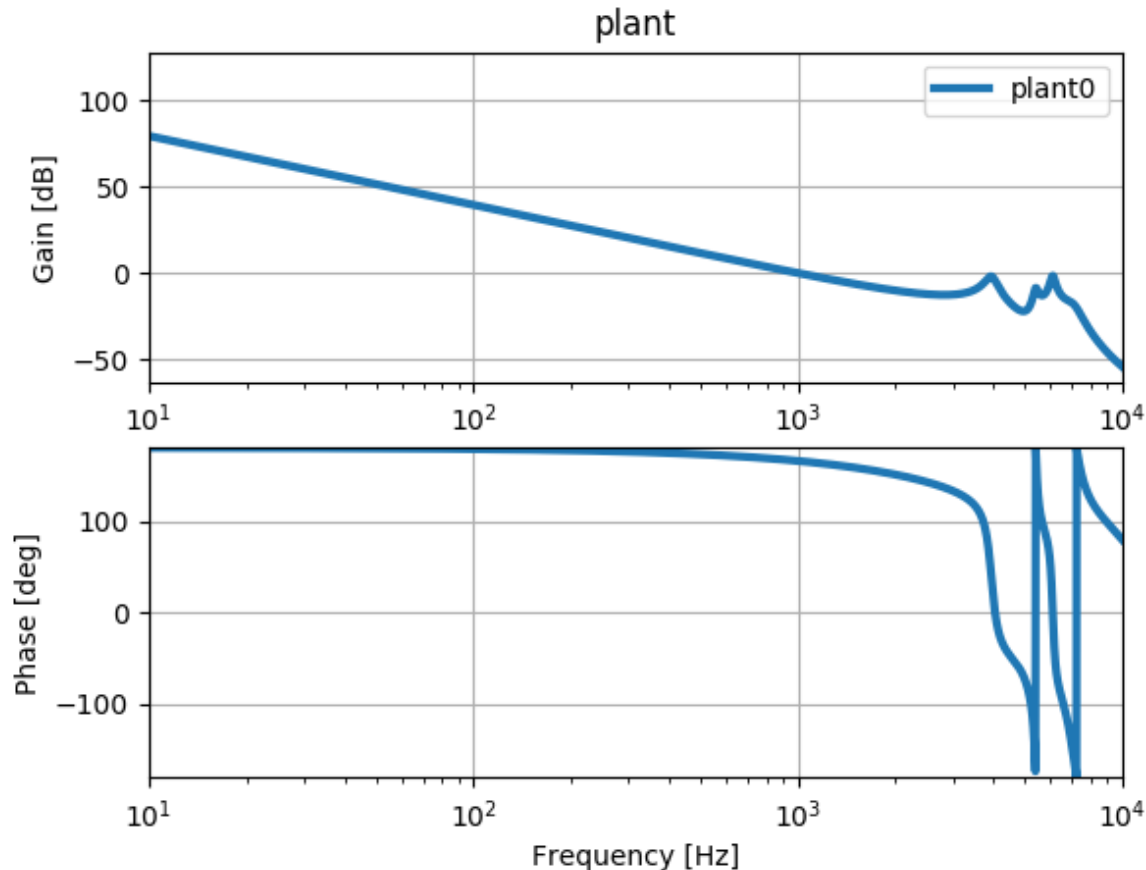
- 20 dB/dec when  $m = 1$ ,
- 40 dB/dec when  $m = 2$ ,
- 60 dB/dec when  $m = 3$ ,
- 80 dB/dec when  $m = 4 \dots$

(The CCCP is omitted since this problem is equivalent to 2.1)

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# 3 1 Plant and Controller



Discretization: 50 us  
Delay: 15 us

The controller is P-I-D (pseudo derivative with 100 us cutoff).

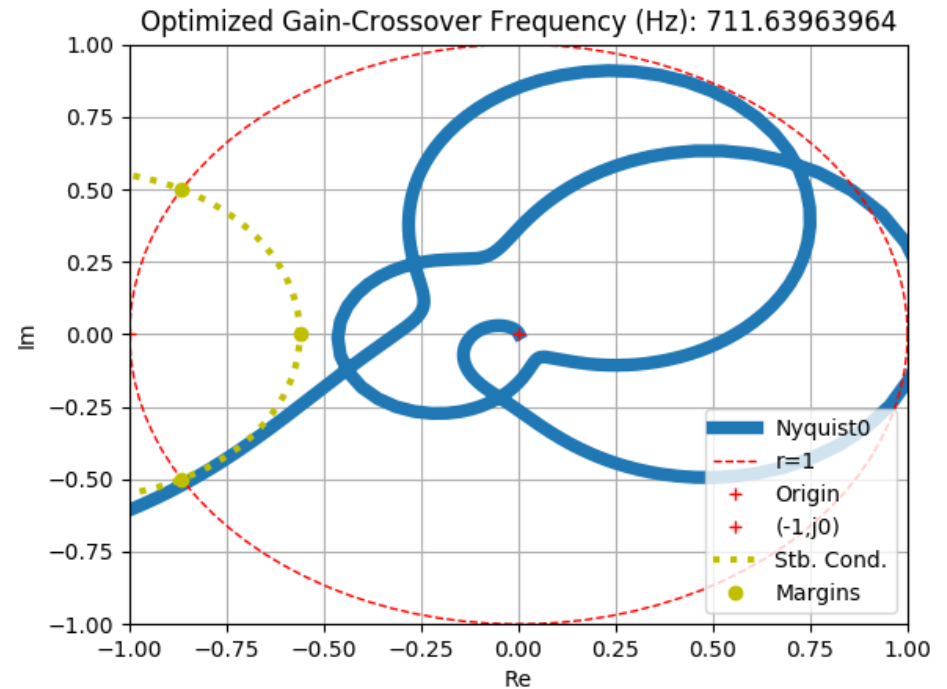
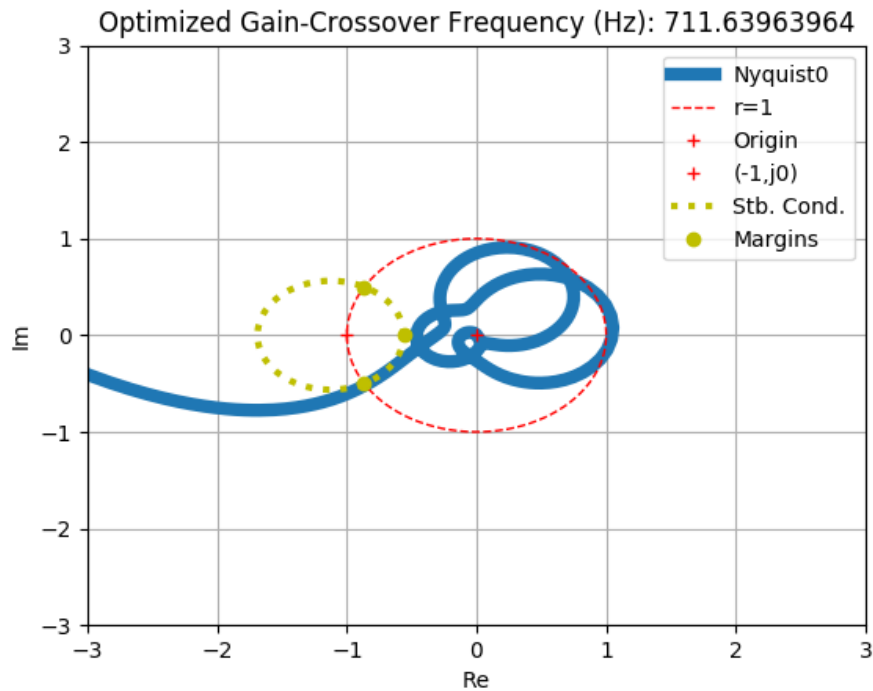
The controller is optimized by iteration of bisection search of  $\omega_{ns}$ .

Gain-Margin: 5 dB, Phase-Margin: 30 degrees, Second Phase-Margin : 30 degrees

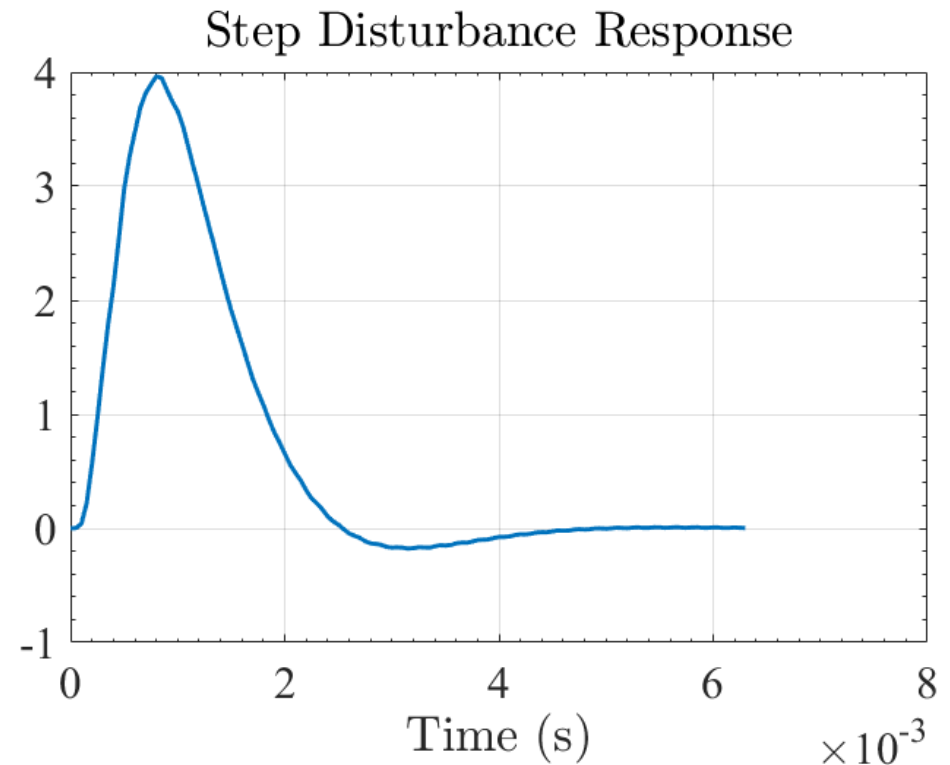
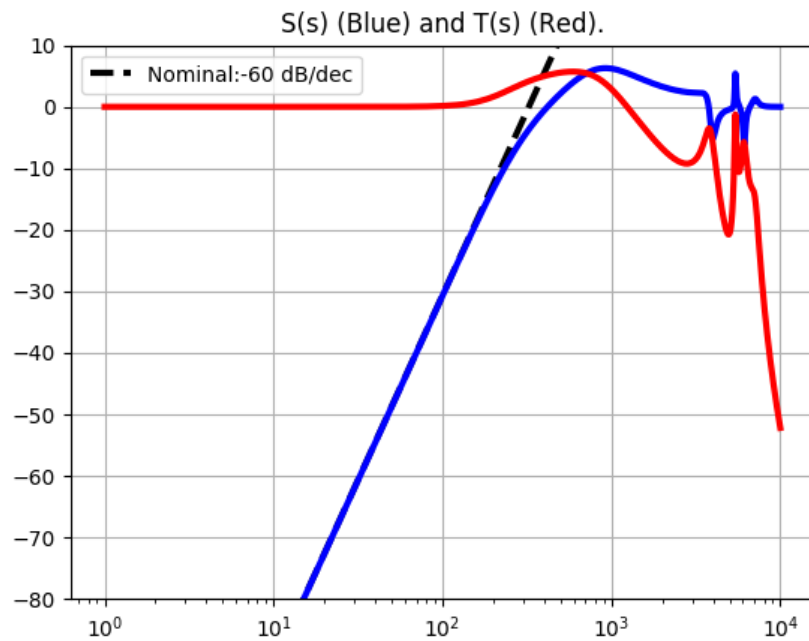
Nominal Sensitivity: -60 dB/dec

## 3.2 Nyquist

P Gain: 2.16684594e-01, I Gain: 2.29603303e+02, D gain: 1.06905746e-04.



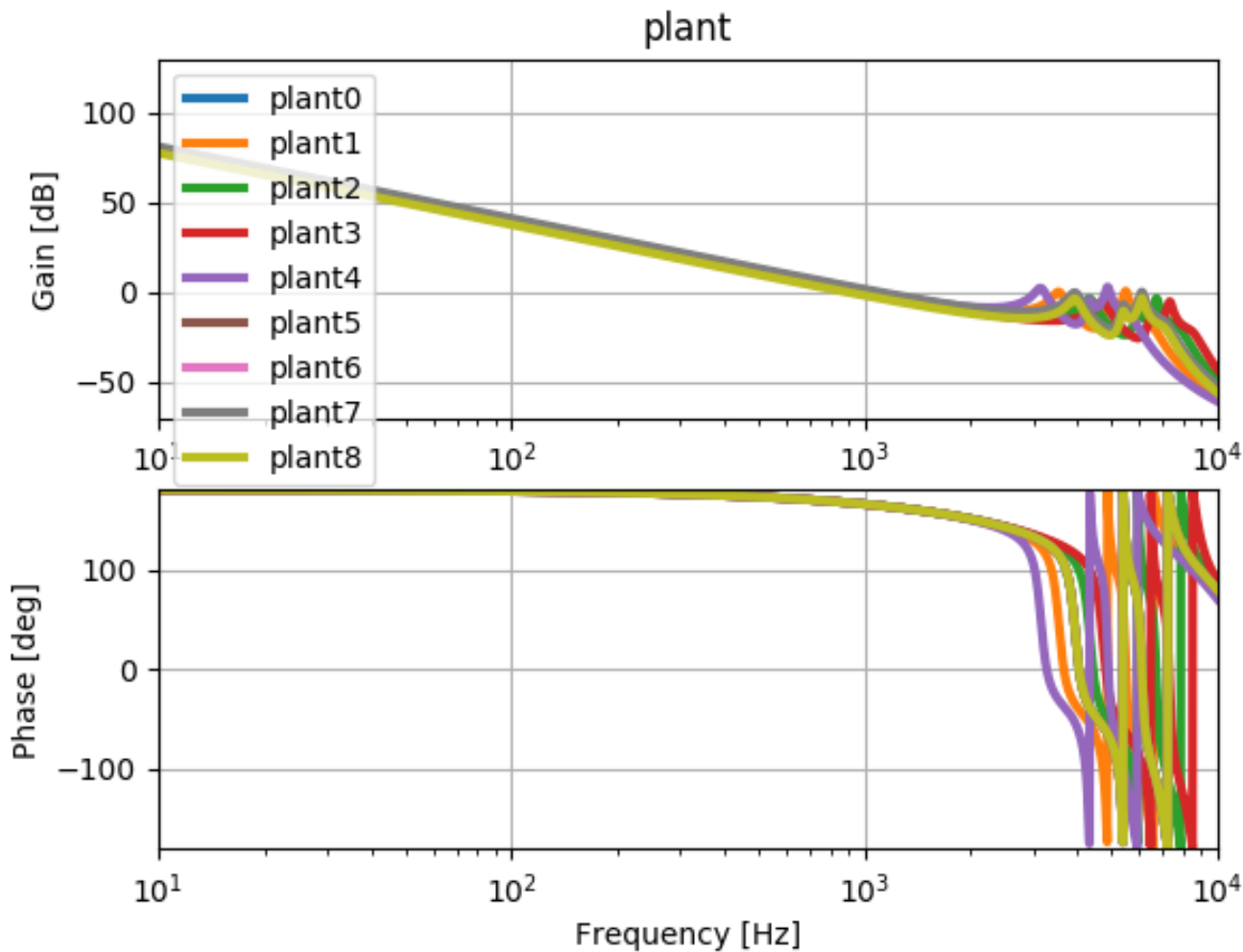
### 3.3 Sensitivity and Step Disturbance Response



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# 4.1 Plant

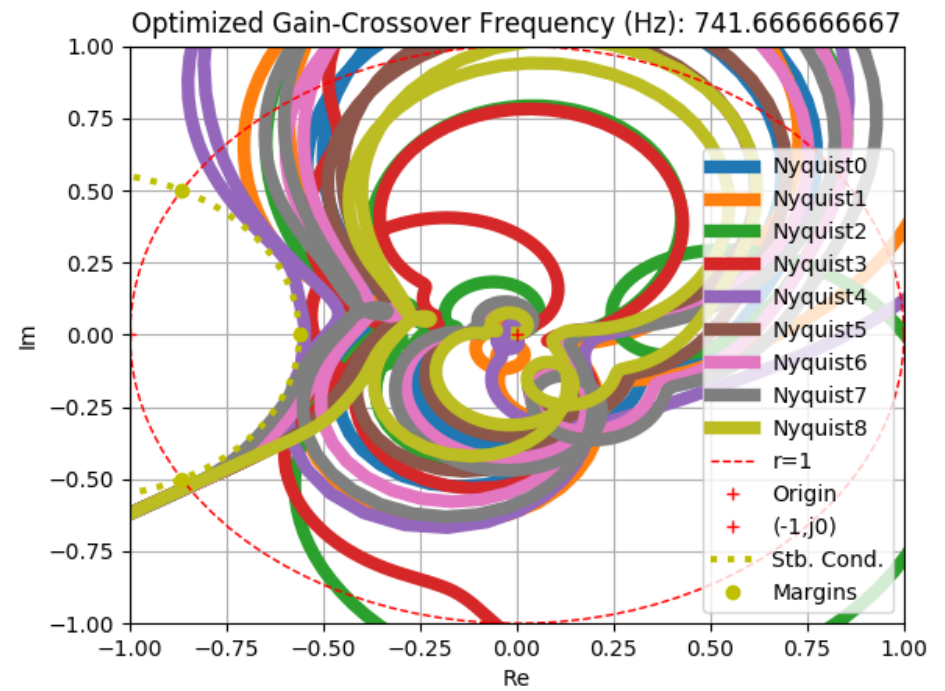
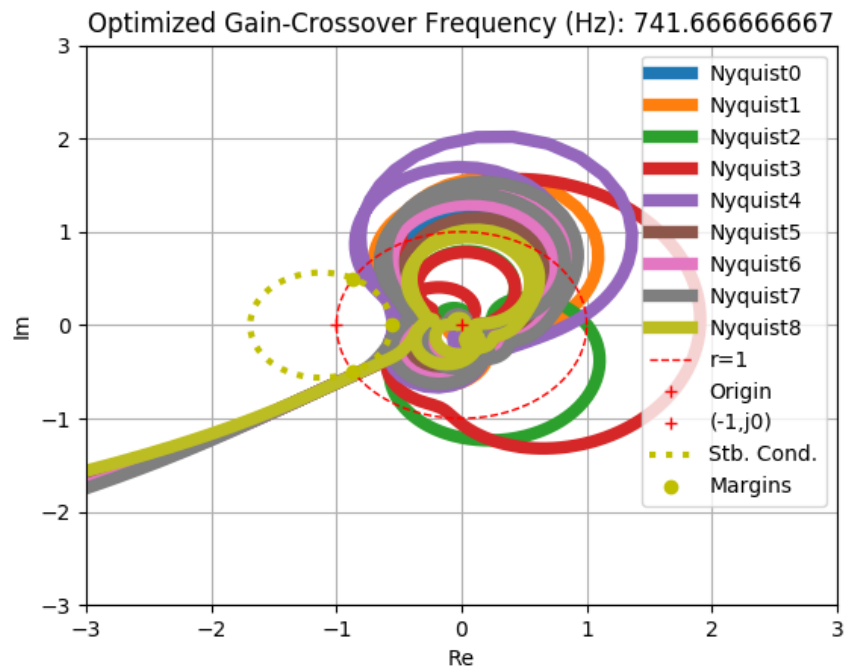


Discretization: 50  $\mu$ s  
 Delay: 15  $\mu$ s

The controller is PID with 30 parallel FIRs  
 Gain-Margin: 5 dB, Phase-Margin: 30 degrees, Second Phase-Margin : 30 degrees  
 Nominal Sensitivity: -40 dB/dec



## 4.2 Nyquist



## 4.3 Sensitivity and Openloop

