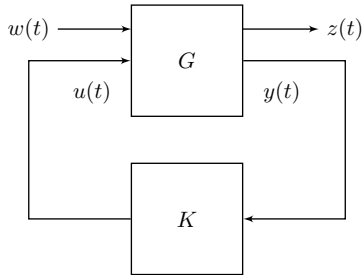


# $\mathcal{H}_2$ Optimal State Feedback Control Synthesis

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# Motivation

# Motivation



- $w(t)$  are exogenous signals – reference, process noise, sensor noise
- $z(t)$  are signals we want to keep small –  $e(t)$ ,  $u(t)$
- Find controller  $K$  that minimizes  $\|G_{w \rightarrow z}\|_2$

- Appropriate when spectral density function of  $w(t)$  is known
  - ▶ For example  $w(t)$  can be stationary noise
  - ▶  $\mathcal{H}_2$  optimal is then linear quadratic Gaussian (LQG)

# Motivation

## Stochastic Input

- For stationary stochastic process  $w(t)$ , **autocorrelation matrix** is

$$R_w(\tau) := \mathbf{E} [w(t + \tau)w^*(t)]$$

- The Fourier transform of  $R_w(\tau)$  is the **spectral density**  $\hat{S}_w(j\omega)$
- Signal power is related to it by

$$\mathbf{E} [|w(t)|]^2 := \text{tr} [R_w(0)] = \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr} [\hat{S}_w(j\omega)] d\omega$$

- If  $z$  and  $w$  are related by  $z = Pw$ , for a **stable LTI** system  $P$ , then

$$\hat{S}_z(j\omega) = \hat{P}(j\omega)\hat{S}_w(j\omega)\hat{P}^*(j\omega)$$

# Motivation

Stochastic Input (contd.)

- Therefore,

$$\mathbf{E} [|z(t)|]^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr} \left[ \hat{P}(j\omega) \hat{S}_w(j\omega) \hat{P}^*(j\omega) \right] d\omega.$$

Looks like weighted  $\mathcal{H}_2$  norm of  $P$ .

- If  $w(t)$  is white noise  $\implies \hat{S}_w(j\omega) = I$ .  
 $\mathcal{H}_2$  norm is output variance with white noise input.
- For any other  $\hat{S}_w(j\omega)$ ,

$$\hat{S}_w(j\omega) = \hat{W}(j\omega) \hat{W}^*(j\omega).$$

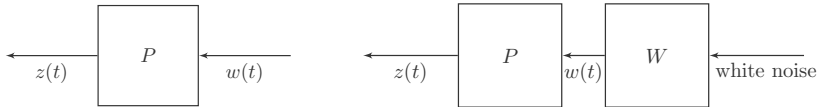
Therefore,

$$\mathbf{E} [|z(t)|]^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr} \left[ \left( \hat{P}(j\omega) \hat{W}(j\omega) \right) \mathbf{I} \left( \hat{W}^*(j\omega) \hat{P}^*(j\omega) \right) \right] d\omega.$$

- Think of  $\hat{P}(j\omega) \hat{W}(j\omega)$  as **weighted system**

# Motivation

Stochastic Input (contd.)



$$\mathbf{E} [|z(t)|]^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr} \left[ \left( \hat{P}(j\omega) \hat{W}(j\omega) \right) \mathbf{I} \left( \hat{W}^*(j\omega) \hat{P}^*(j\omega) \right) \right] d\omega.$$

Think of  $\hat{P}(j\omega) \hat{W}(j\omega)$  as **weighted system**

# Motivation

## *Impulse Response*

- Input signal is known in advance
- Tracking a fixed signal – e.g. step

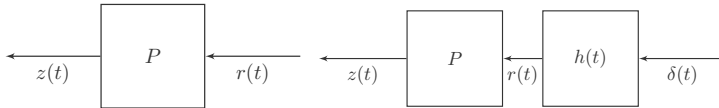
Consider a special case: Scalar  $w(t) = \delta(t)$ . Implies

$$\begin{aligned}\|z\|_2^2 &= \int_0^\infty z^*(t)z(t)dt \\ &= \frac{1}{2\pi} \int_{-\infty}^\infty \hat{z}^*(j\omega)\hat{z}(j\omega)d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^\infty \hat{P}^*(j\omega)\hat{P}(j\omega)d\omega \\ &= \|\hat{P}\|_2^2\end{aligned}$$

# Motivation

*Impulse Response (contd.)*

What about any other reference  $r(t)$ ?



- $r(t)$  can be replaced by the impulse response  $h(t)$  of a known filter  $W(j\omega)$
- $z(t)$  becomes impulse response of a weighted plant



# Computation of $\mathcal{H}_2$ Norm

- Best computed in state-space realization of system

**State Space Model:** General MIMO LTI system modeled as

$$\dot{x} = Ax + Bu,$$

$$y = Cx + Du,$$

where  $x \in \mathbb{R}^n, u \in \mathbb{R}^m, y \in \mathbb{R}^p$ .

## Transfer Function

$$\hat{G}(s) = D + C(sI - A)^{-1}B \text{ strictly proper when } D = 0$$

## Impulse Response

$$G(t) = \mathcal{L}^{-1} \{C(sI - A)^{-1}B\} = Ce^{tA}B.$$

# $\mathcal{H}_2$ Norm

## MIMO Systems

$$\begin{aligned}\|\hat{G}(j\omega)\|_2^2 &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{tr} \left[ \hat{G}^*(j\omega) \hat{G}(j\omega) \right] \text{ for matrix transfer function} \\ &= \|G(t)\|_2^2 \text{ Parseval} \\ &= \int_0^{\infty} \text{tr} \left[ C e^{tA} B B^T e^{tA^T} C^T \right] dt \\ &= \text{tr} \left[ C \underbrace{\left( \int_0^{\infty} e^{tA} B B^T e^{tA^T} dt \right)}_{L_c} C^T \right] \quad L_c = \text{controllability Gramian} \\ &= \text{tr} [C L_c C^T]\end{aligned}$$

# $\mathcal{H}_2$ Norm (contd.)

MIMO Systems

For any matrix  $M$

$$\begin{aligned}\text{tr} [M^* M] &= \text{tr} [M M^*] \\ \Rightarrow \|\hat{G}(j\omega)\|_2^2 &= \text{tr} \left[ B^T \underbrace{\left( \int_0^\infty e^{tA^T} C^T C e^{tA} dt \right)}_{L_o} B \right] \\ &= \text{tr} [B^T L_o B] \quad L_o = \text{observability Gramian}\end{aligned}$$

## $\mathcal{H}_2$ Norm of $\hat{G}(j\omega)$

$$\|\hat{G}(j\omega)\|_2^2 = \text{tr} [C L_c C^T] = \text{tr} [B^T L_o B] .$$

# $\mathcal{H}_2$ Norm

How to determine  $L_c$  and  $L_o$ ?

They are solutions of the following equation

$$AL_c + L_cA^T + BB^T = 0,$$

$$A^TL_o + L_oA + C^TC = 0.$$

# State-Feedback $\mathcal{H}_2$ Synthesis

**Proposition 1** Suppose  $P$  is a state-space system with realization  $(A, B, C)$ . Then

$$A \text{ is Hurwitz and } \|\hat{P}\|_2 \leq 1,$$

iff  $\exists X = X^T > 0$  such that

$$\text{tr}[CXC^*] < 1 \text{ and } AX + XA^* + BB^* < 0.$$

**Proof Only If**

Recall  $\|\hat{P}\|_2 = \text{tr}[CL_cC^*]$ . Therefore for  $X = L_c$

$$\text{tr}[CL_cC^*] < 1 \implies \|\hat{P}\|_2 < 1,$$

and  $A$  is Hurwitz  $\|\hat{P}\|_2$  is finite

# State-Feedback $\mathcal{H}_2$ Synthesis

contd.

Consider

$$X = \int_0^{\infty} e^{tA}(BB^* + \epsilon I_n)e^{tA^*} dt,$$

- is continuous in  $\epsilon$
- equals to  $L_c$  when  $\epsilon = 0$ .

It can be shown that this  $X$  satisfies Lyapunov equation follow the derivation

$$AX + XA^* + BB^* + \epsilon I_n = 0,$$

or

$$AX + XA^* + BB^* < 0.$$

# State-Feedback $\mathcal{H}_2$ Synthesis

contd.

**Proof** If

$X$  satisfies

$$AX + XA^* + BB^* < 0$$

and

$$\text{tr}[CXC^*] < 1.$$

Implies,  $A$  is Hurwitz.

**Proposition:** It can be shown that if  $L_c$  satisfies

$$A^*L_c + L_cA + Q = 0,$$

and  $X$  satisfies

$$A^*X + XA + Q \leq 0,$$

then  $X \geq L_c$ .

# State-Feedback $\mathcal{H}_2$ Synthesis

*contd.*

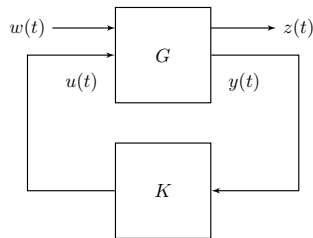
Inequality  $X \geq L_c$  implies

$$\|\hat{P}\|_2 = \text{tr} [CL_c C^*] \leq \text{tr} [CXC^*] < 1.$$



# State-Feedback $\mathcal{H}_2$ Synthesis

*System Dynamics*



## Dynamics

$$\begin{aligned} \dot{x} &= Ax + B_w w + B_u u \\ z &= C_z x + D_u u \\ y &= x \end{aligned} \quad \Rightarrow \quad \hat{G}(s) = \left[ \begin{array}{c|cc} A & B_w & B_u \\ \hline C_z & 0 & D_u \\ I & 0 & 0 \end{array} \right]$$

With  $u = Kx$ ,

$$\mathcal{F}_l(\hat{G}, K) := \hat{G}_{w \rightarrow z}(s) = \left[ \begin{array}{c|c} A + B_u K & B_w \\ \hline C_z + D_u K & 0 \end{array} \right].$$

# State-Feedback $\mathcal{H}_2$ Synthesis

## Controller Synthesis

**Proposition** There exists feedback gain  $K$  that internally stabilizes  $G$  and satisfies

$$\|\mathcal{F}_l(\hat{G}, K)\|_2 < 1$$

iff  $\exists Z \in \mathbb{R}^{m \times n}$  such that

$$K = ZX^{-1},$$

where  $X > 0$  satisfies inequalities

$$\begin{bmatrix} A & B_u \end{bmatrix} \begin{bmatrix} X \\ Z \end{bmatrix} + \begin{bmatrix} X & Z^* \end{bmatrix} \begin{bmatrix} A^* \\ B_u^* \end{bmatrix} + B_w B_w^* < 0,$$
$$\text{tr} [(C_z X + D_u Z) X^{-1} (C_z X + D_u Z)^*] < 1.$$

**Proof:** Use closed-loop state-space data and earlier proposition.

Not an LMI.

# State-Feedback $\mathcal{H}_2$ Synthesis

## Controller Synthesis (contd.)

Apply Schur complement to get following convex problem.

There exists feedback gain  $K$  that internally stabilizes  $G$  satisfying

$$\|\mathcal{F}_l(\hat{G}, K)\|_2 < 1$$

iff  $\exists X \in \mathbb{R}^{n \times n}$ ,  $W \in \mathbb{R}^{q \times q}$ , and  $Z \in \mathbb{R}^{m \times n}$ , such that

$$K = ZX^{-1},$$

and

$$\begin{aligned} & \min_W \text{tr}[W] \\ & \begin{bmatrix} A & B_u \end{bmatrix} \begin{bmatrix} X \\ Z \end{bmatrix} + \begin{bmatrix} X & Z^* \end{bmatrix} \begin{bmatrix} A^* \\ B_u^* \end{bmatrix} + B_w B_w^* < 0, \\ & \begin{bmatrix} W & (C_z X + D_u Z) \\ (C_z X + D_u Z)^* & X \end{bmatrix} > 0 \end{aligned}$$

# State-Feedback $\mathcal{H}_2$ Synthesis

*Very simple example*

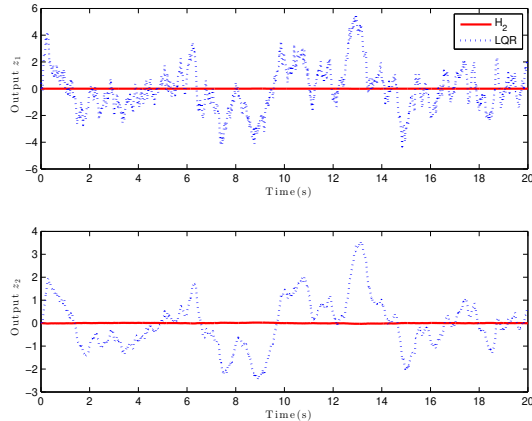
## System Dynamics

$$\dot{x} = \begin{bmatrix} -3 & -2 & 1 \\ 1 & 2 & 1 \\ 1 & -1 & -1 \end{bmatrix} x + \begin{bmatrix} 2 & 0 \\ 0 & 2 \\ 0 & 1 \end{bmatrix} u + \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix} w,$$
$$z = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} x + \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} u.$$

Compare with LQR

# State-Feedback $\mathcal{H}_2$ Synthesis

*Very simple example*



$\text{tr}[W^*] = 3.3491e - 4$  – Good disturbance attenuation.

# MATLAB Code

```
clear; clc;

A = [-3 -2 1;
      1 2 1;
      1 -1 -1;];
Bu = [2 0; 0 2; 0 1];
Bw = [3;0;1];

Cz = [1 0 1; 0 1 1];
Du = [1 1; 0 1];

nx = 3;
nu = 2;
nz = 2;
nw = 1;

cvx_begin sdp
    variable X(nx,nx) symmetric
    variable W(nz,nz) symmetric
    variable Z(nu,nx)

    [A Bu]*[X;Z] + [X Z']*[A';Bu'] + Bw*Bw' < 0
    [W (Cz*X + Du*Z); (Cz*X + Du*Z)' X] > 0
    minimize trace(W)
cvx_end

h2K = Z*inv(X);
[lqrK,S,E] = lqr(A,Bu,Cz'*Cz,Du'*Du);

h2G = ss(A+Bu*h2K, Bw, Cz + Du*h2K, zeros(nw,1));
lqrG = ss(A-Bu*lqrK, Bw, Cz + Du*lqrK, ...
          zeros(nw,1));

T = [0:0.01:20];
w = 5*randn(length(T),1);
[y1,t1,x1] = lsim(h2G,w,T);
[y2,t2,x2] = lsim(lqrG,w,T);
f1=figure(1); clf;
set(f1,'defaulttextinterpreter','latex');

for i=1:2
    subplot(2,1,i);
    plot(t1,y1(:,i),'r',t2,y2(:,i),'b:', ...
         'linewidth',2);
    xlabel('Time(s)');
    ylabel(sprintf('Output %z_d$',i));
end
subplot(2,1,1); legend('H_2','LQR');
print -depsc h2ex1.eps
```

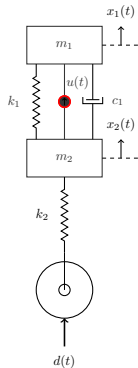
# State-Feedback $\mathcal{H}_2$ Synthesis

*Regulator with disturbance – Active Suspension*

## Equations of motion

$$m_1 \ddot{x}_1 = -k_1(x_1 - x_2) - c_1(\dot{x}_1 - \dot{x}_2) + u,$$

$$m_2 \ddot{x}_2 = k_1(x_1 - x_2) + c_1(\dot{x}_1 - \dot{x}_2) - k_2(x_2 - d) - u.$$



## State Variables

$$q_1 := x_1,$$

$$q_2 := x_2,$$

$$q_3 := \dot{x}_1,$$

$$q_4 := \dot{x}_2.$$

# State-Feedback $\mathcal{H}_2$ Synthesis

*Regulator with disturbance – Active Suspension*

## Linear System

$$\begin{bmatrix} \dot{q}_1 \\ \dot{q}_2 \\ \dot{q}_3 \\ \dot{q}_4 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -k_1/m_1 & k_1/m_1 & -c_1/m_1 & c_1/m_1 \\ k_1/m_2 & -(k_1 + k_2)/m_2 & c_1/m_2 & -c_1/m_2 \end{bmatrix} \begin{bmatrix} q_1 \\ q_2 \\ q_3 \\ q_4 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1/m_1 \\ -1/m_2 \end{bmatrix} u + \begin{bmatrix} 0 \\ 0 \\ 0 \\ k_2/m_2 \end{bmatrix} d.$$

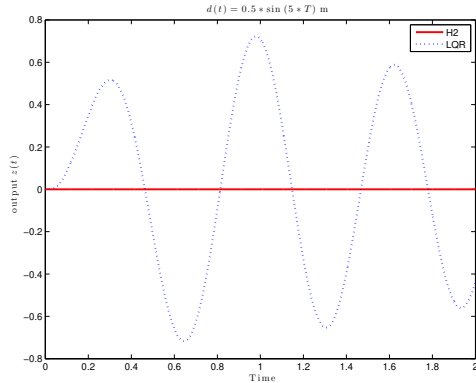
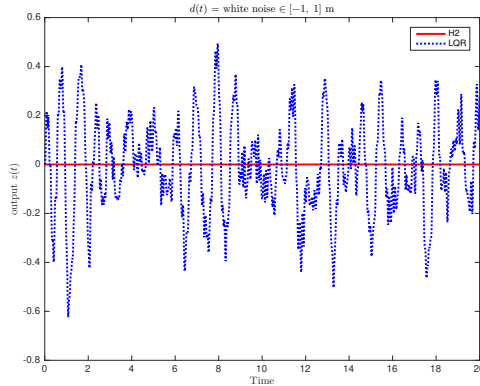
## Output $z(t)$

$$z = q_1 + u.$$



# State-Feedback $\mathcal{H}_2$ Synthesis

*Regulator with disturbance – Active Suspension*



$\text{tr}[W^*] = 1.17367e - 10$  – Very good disturbance rejection.

# MATLAB Code

```
clear; clc;
% System Parameters
m1 = 290; % kg -- Body mass
m2 = 60; % kg -- suspension mass
```

```
k1 = 16200; % N/m
k2 = 191000; % N/m
c1 = 1000; % Ns/m
```

```
A = [0 0 1 0;
      0 0 0 1;
      -k1/m1 k1/m1 -c1/m1 c1/m1;
      k1/m1 -(k1+k2)/m2 c1/m2 -c1/m2];
Bu = [0;0;1/m1;-1/m2];
Bw = [0;0;0;k2/m2];
```

```
nx = 4; nu = 1;
nz = 1; nw = 1;
```

```
Cz = [1,0,0,0];
Du = 1*ones(nz,nu);
```

```
cvx_begin sdp
    variable X(nx,nx) symmetric
    variable W(nz,nz) symmetric
    variables Z(nu,nx) gam

    [A Bu]*[X;Z] + [X Z']*[A';Bu'] + Bw*Bw' < 0
    W (Cz*X + Du*Z); (Cz*X + Du*Z)' X > 0
    minimize trace(W)
cvx_end

h2K = Z*inv(X);
[lqrK,S,E] = lqr(A,Bu,Cz'*Cz,Du'*Du);

% Simulation
h2G = ss(A+Bu*h2K, Bw, Cz + Du*h2K, zeros(nz,nw));
lqrG = ss(A-Bu*lqrK, Bw, Cz + Du*lqrK, zeros(nz,nw));
T = [0:0.01:20]/10;
w = 2*rand(length(T),1)-1;
w = 0.5*sin(10*T);
[y1,t1,x1] = lsim(h2G,w,T);
[y2,t2,x2] = lsim(lqrG,w,T);

f1 = figure(1); clf;
set(f1,'defaulttextinterpreter','latex');
plot(t1,y1,'r',t2,y2,'b','Linewidth',2);
xlabel('Time'); ylabel('output $z(t)$');
title('$d(t) = 0.5*\sin\; ; (5*T)$ m');
title('$d(t)$ = white noise $\in [-1,1]$ m');
legend('H2','LQR');
print -depsc h2qcar2.eps
```

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance*

Let dynamical system be

$$\dot{x} = Ax + B_u u + B_d d, \quad z = \begin{bmatrix} x \\ u \end{bmatrix}, \quad y = C_y x.$$

- Design a controller of the form

$$u = Kx + Gr$$

- $K$  is designed in  $\mathcal{H}_2$  optimal sense
- $G$  is regular tracking gain

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance*

Closed-loop system with  $K$  is therefore

$$\dot{x} = (A + BK)x + B_u Gr, \quad y = C_y x.$$

- Ignore disturbance when determining  $G$
- Steady-state response to **constant**  $r$  is

$$0 = (A + BK)x_{ss} + BGr, \quad y_{ss} = C_y x_{ss} = r.$$

Or  $x_{ss} = -(A + BK)^{-1}BGr$ , implies

$$-C_y(A + BK)^{-1}BGr = r,$$

or

$$-C_y(A + BK)^{-1}BG = I.$$

- Solve for  $G$

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance*

- Existence of solution of

$$C_y(A + BK)^{-1}BG = I$$

is **necessary and sufficient condition** for existence of a tracking controller.

- In general, when

$$y = C_yx + D_yu,$$

the equation becomes

$$[D_y - (C_y + D_yK)(A + BK)^{-1}B] G = I.$$

- Can be rewritten in terms of  **$\Pi$**

$$\Pi = -(A + BK)^{-1}BG \implies (C_y + D_yK)\Pi + D_yG = I,$$

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance*

Or

$$\begin{aligned}(A + BK)\Pi + BG &= 0, \\ (C_y + D_y K)\Pi + D_y G &= I.\end{aligned}$$

Or rearranged to

$$\begin{aligned}A\Pi + B(K\Pi + G) &= 0, \\ C_y\Pi + D_y(K\Pi + G) &= I.\end{aligned}$$

Therefore, with  $\Gamma := K\Pi + G$ , we get

$$\begin{bmatrix} A & B \\ C_y & D_y \end{bmatrix} \begin{bmatrix} \Pi \\ \Gamma \end{bmatrix} + \begin{bmatrix} 0_{n_x \times n_y} \\ -I_{n_y \times n_y} \end{bmatrix} = 0.$$

This is the so called **regulator equation**. Get  $G = \Gamma - K\Pi$ .

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance*

- Solve regulator equation

$$\begin{bmatrix} A & B \\ C_y & D_y \end{bmatrix} \begin{bmatrix} \Pi \\ \Gamma \end{bmatrix} + \begin{bmatrix} 0_{n_x \times n_y} \\ -I_{n_y \times n_y} \end{bmatrix} = 0.$$

- Get  $G = \Gamma - K\Pi$ .
- Control law is

$$u = Kx + Gr.$$

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance – Longitudinal F16 Control Law*

## Longitudinal Motion

- States  $[V(\text{ft/s}) \ \alpha(\text{rad}) \ \theta(\text{rad}) \ q(\text{rad/s})]^T$
- Controls  $[T(\text{lb}) \ \delta_e(\text{deg})]$
- Constraints on  $u$

$$\begin{bmatrix} 1000 \\ -25 \end{bmatrix} \leq u \leq \begin{bmatrix} 19000 \\ 25 \end{bmatrix}$$



- Trimmed at steady-level flight at 932 ft/s

$$x_{\text{trim}} = \begin{bmatrix} 932.2894 \\ 0 \\ 0 \\ 0 \end{bmatrix}, u_{\text{trim}} = \begin{bmatrix} 5318.2 \\ -1.3935 \end{bmatrix}$$

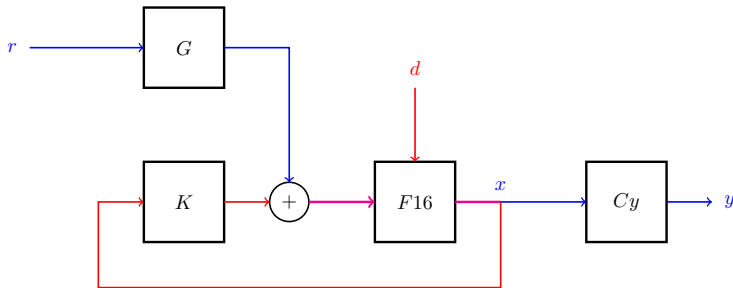
- Disturbance in  $\dot{\alpha}$  equation



# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance – Longitudinal F16 Control Law*

## System Interconnection



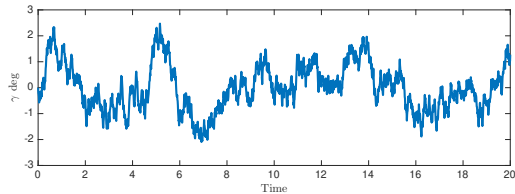
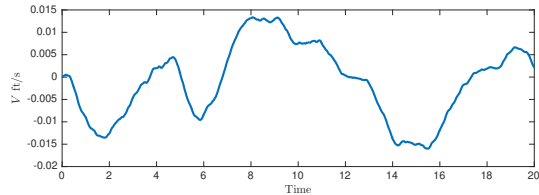
## Control Objective

- Design  $K$  to minimize  $\|G_{d \rightarrow x}\|_2$
- Design  $G$  to track  $r := \begin{bmatrix} V \\ \gamma \end{bmatrix}$  reference

# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance – Longitudinal F16 Control Law*

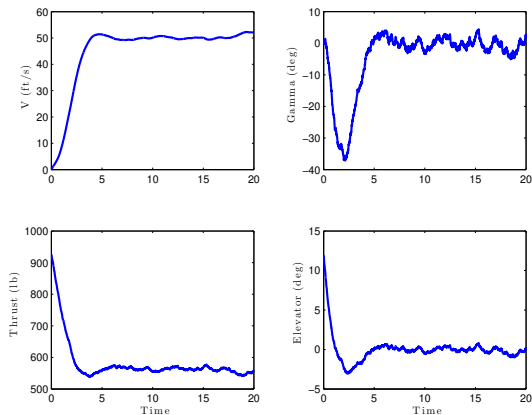
**Disturbance Rejection**  $d(t) \in \mathcal{U}_{[-1,1]}$  rad,  $\text{tr}[W^*] = 0.118159$



# State-Feedback $\mathcal{H}_2$ Synthesis

Tracking with disturbance – Longitudinal F16 Control Law

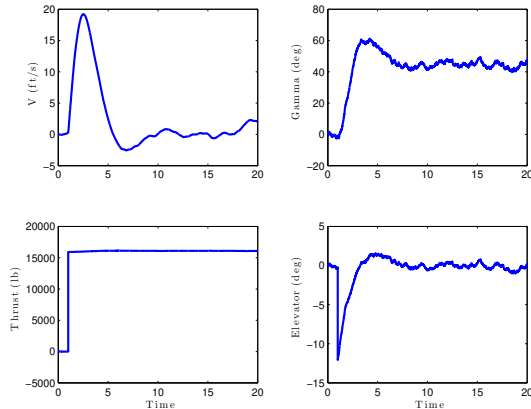
Tracking Performance  $V_{\text{ref}} = 50 \text{ ft/s step}$ ,  $\gamma_{\text{ref}} = 0$



# State-Feedback $\mathcal{H}_2$ Synthesis

Tracking with disturbance – Longitudinal F16 Control Law

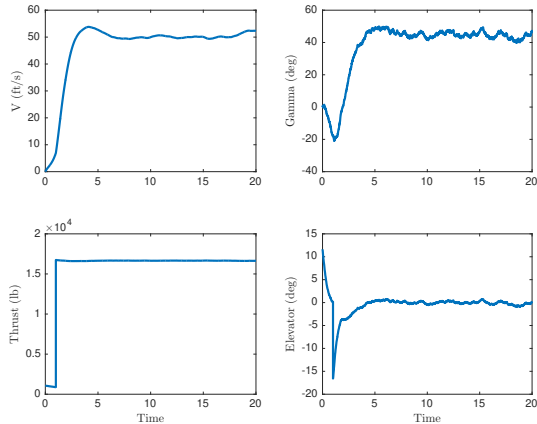
Tracking Performance  $V_{\text{ref}} = 0$ ,  $\gamma_{\text{ref}} = 45$  deg step



# State-Feedback $\mathcal{H}_2$ Synthesis

Tracking with disturbance – Longitudinal F16 Control Law

Tracking Performance  $V_{\text{ref}} = 50 \text{ ft/s step}$ ,  $\gamma_{\text{ref}} = 45 \text{ deg step}$



# State-Feedback $\mathcal{H}_2$ Synthesis

*Tracking with disturbance – Longitudinal F16 Control Law*

## Simulink

