

# Robust Multivariable Control

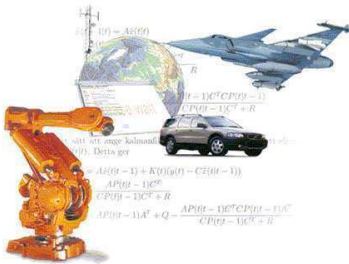
## Lecture 1

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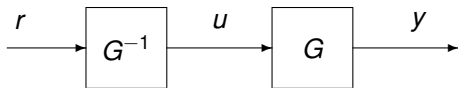
<https://users.isy.liu.se/rt/andersh/teaching/robkurs.html>

<https://users.isy.liu.se/rt/andersh/teaching/robschedule.html>



# Feedback

Why do we need feedback?



We want the output  $y$  to follow the reference input  $r$ .

When *cannot* we do like this?

$G$  is unstable;

$G$  is uncertain;

$G$  is not minimum phase (zeros in RHP,  $G^{-1}$  unstable).



# A failure – Conestoga 1620 rocket failure

[https://www.youtube.com/watch?v=wWDJBkf\\_P3Yi](https://www.youtube.com/watch?v=wWDJBkf_P3Yi)



# A success – SpaceX core stage landing

[https://youtu.be/sYmQQn\\_ZSys](https://youtu.be/sYmQQn_ZSys)

Simple dynamics

$$\frac{d}{dt} \begin{bmatrix} \dot{\theta} \\ \theta \\ \dot{y} \\ y \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & T/m & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{\theta} \\ \theta \\ \dot{y} \\ y \end{bmatrix} + \begin{bmatrix} -\ell T/l \\ 0 \\ T/m \\ 0 \end{bmatrix} \delta$$

Attitude angle,  $\theta$ , and lateral position,  $y$ , as outputs.

Nozzle deflection,  $\delta$ , as input.

Essentially the same dynamics for a Lunar landing: two double integrators.

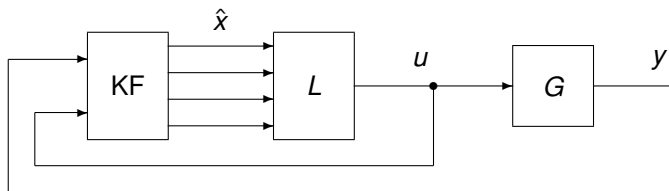


# An example – SpaceX core stage landing

- 1 The same dynamics applies to both liftoff and landing.
- 2 What are the differences ?
- 3 Two, three or four integrators in series ?
- 4 How does this affect the robustness margin?
- 5 How can we handle several inputs and outputs?



# Some history



LQR/LQE is optimal in some sense

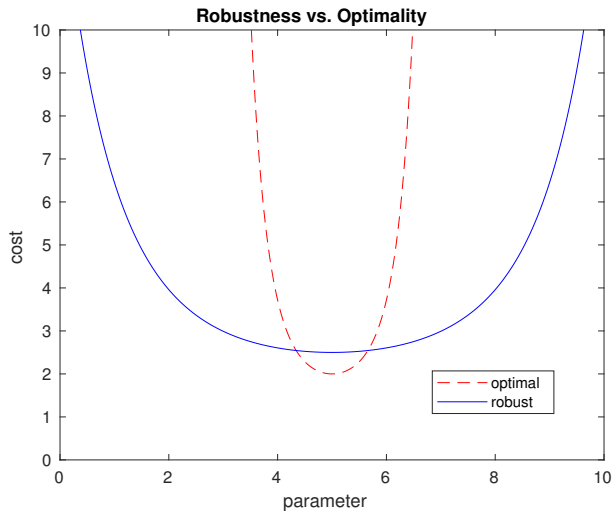
What about phase and amplitude margins?

Abstract: “There are none”, Doyle, 1978, chapter 14.10 in ZDG.

LTR gives margins (Loop Transfer Recovery).

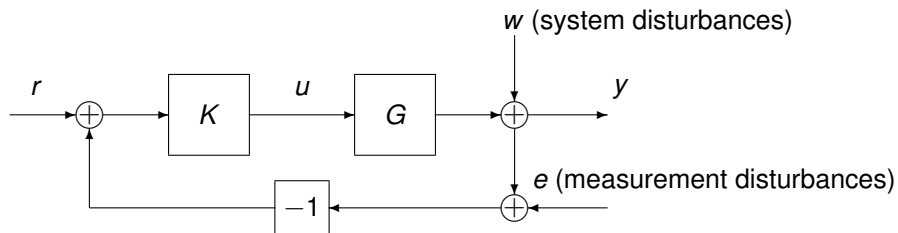


# Robustness





# Standard form



$$y = Sw + T(r - e)$$

$$S = (I + GK)^{-1}$$

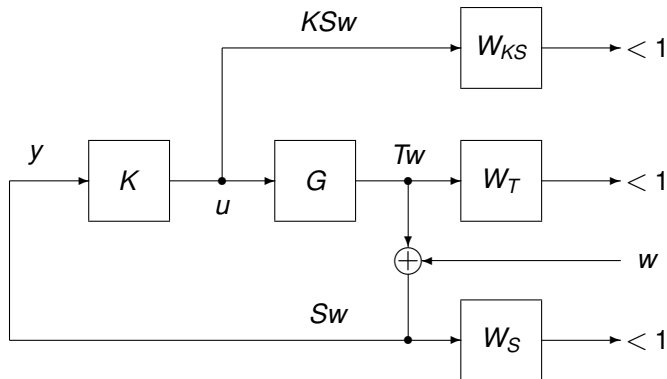
$$T = GK(I + GK)^{-1}$$

$$S + T = (I + GK)(I + GK)^{-1} = I$$

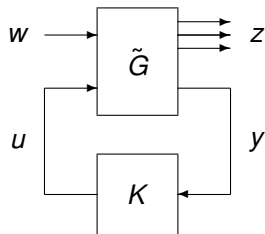


# Augmented system

We would also like to reduce the control signals in magnitude.



# Augmented system

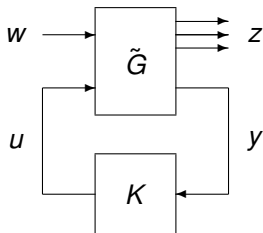


We want to limit the gain from  $w$  to  $z$  by a suitable choice of  $K$ .

Try to reduce the gain to one or below.



# Design methods

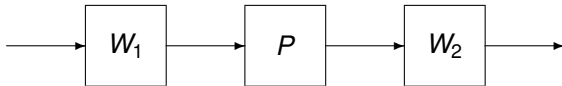


- 1) Specify the requirements
- 2) Synthesis
- 3) Check if the requirements are satisfied
- 4) Modify and repeat from 1)

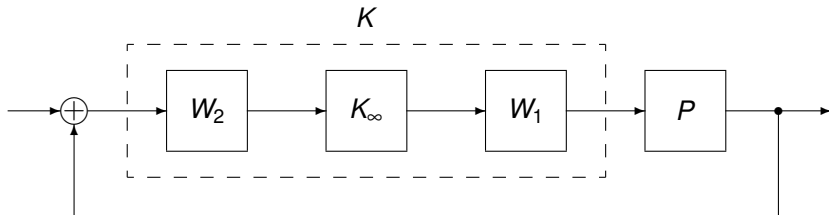


# Loop shaping

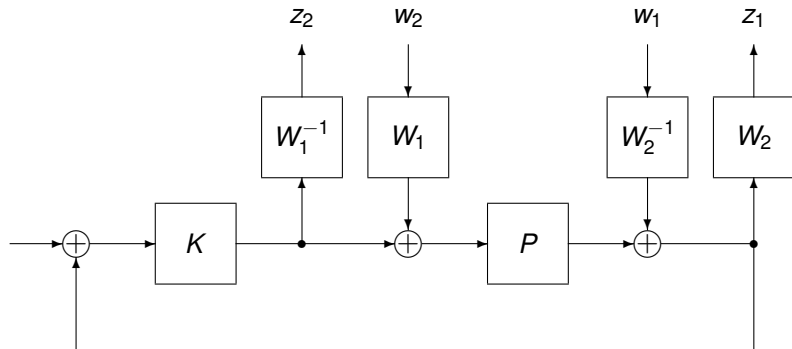
Find  $W_1$  and  $W_2$  to shape the open loop gain



The controller,  $K$ , can be obtained in a synthesis step



# Alternative formulation



# Multivariable gain

Consider

$$z = Mw$$
$$\begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} 0.96 & 1.72 \\ 2.28 & 0.96 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

Let  $\|w\|^2 = w_1^2 + w_2^2 = 1$ , and maximize  $\|z\|^2 = z_1^2 + z_2^2$ .



# Multivariable gain

Singular value decomposition:

$$M = \begin{bmatrix} 0.96 & 1.72 \\ 2.28 & 0.96 \end{bmatrix} = U \Sigma V^T$$
$$= \begin{bmatrix} 0.6 & -0.8 \\ 0.8 & 0.6 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0.8 & 0.6 \\ 0.6 & -0.8 \end{bmatrix}^T$$

where  $U^T U = I$ ,  $V^T V = I$ ,  $\Sigma = \begin{bmatrix} \sigma_1 & & \\ & \ddots & \\ & & \sigma_n \end{bmatrix} \succeq 0$ .

Use `svd` in Matlab.

Maximum gain,  $\|M\| = \bar{\sigma}(M) = \bar{\sigma}(M^T) = \sigma_1 = \max_i \sigma_i$ .

Different directions in  $w$  give different gains.





# Connection to eigenvalues

$$M = U\Sigma V^T$$
$$M^T M = V\Sigma U^T U\Sigma V^T = V\Sigma^2 V^T$$

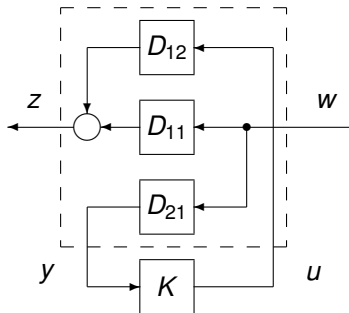
Thus,  $M^T M V = V\Sigma^2$  and  $\Sigma^2$  are eigenvalues to  $M^T M$ .

Also, the eigenvalues of  $\begin{bmatrix} 0 & M \\ M^T & 0 \end{bmatrix}$  are equal to  $\pm\sigma_i$ .

Computing the eigenvalues can be numerically difficult



Consider a static system:



Choose a  $K$  such that  $\|D_{11} + D_{12}KD_{21}\|$  is minimized.



Use orthonormal transformations to obtain a similar problem

$$\text{Find } \gamma = \min_K \left\| \begin{bmatrix} C & K \\ D & B \end{bmatrix} \right\|.$$

This is solved by Parrott's theorem:

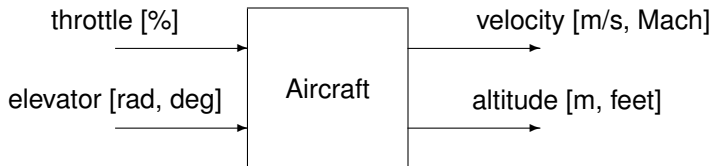
$$\gamma = \max \left\{ \left\| \begin{bmatrix} D & B \end{bmatrix} \right\|, \left\| \begin{bmatrix} C \\ D \end{bmatrix} \right\| \right\}$$

$$\text{One solution is } K = \arg \min_K \text{rank} \begin{bmatrix} & C & K \\ -\gamma & D & B \\ D^T & -\gamma & \end{bmatrix}$$

This problem is related to the elimination lemma.



# Multivariable phenomena



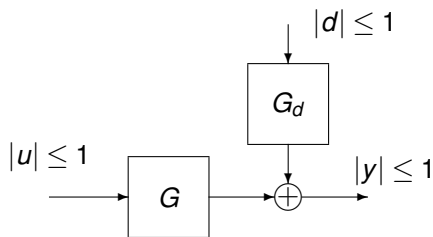
How can we compare rad and %?

How can we compare m, m/s and Mach number?

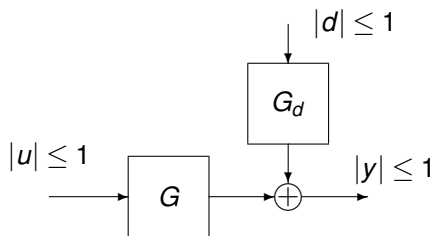


# Scaling

- Scale the input signals,  $u$ , so that they are limited between  $\pm 1$ .
- Scale the output signals,  $y$ , so that they are limited  $\pm 1$ .
- Scale the disturbances,  $d$ , so that they are limited  $\pm 1$ .



# Scaling

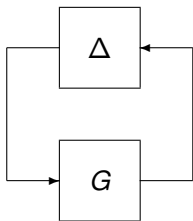


In order to compensate for a disturbance so that  $|y| \leq 1$ :

$$\|G^{-1}G_d\| \leq 1$$



# Small gain theorem



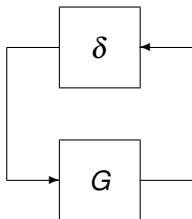
$G$  stable and  $\|G\| < 1$ .

If  $\|\Delta\| \leq 1$  then the closed loop system is stable



# Small gain theorem

Pull out the uncertain parameters.  
LFT = linear fractional transformation:



$$|\delta| \leq 1$$





- 1 Introduction
- 2 Norms, Lyapunov equations, balancing
- 3 Feedback, stability, performance
- 4 Parametrization of regulators, Riccati equations, LQR
- 5  $H_\infty$ -synthesis
- 6 LMIs
- 7 Loop shaping, model reduction
- 8 Model uncertainties, small gain theorem, LFTs
- 9  $\mu$  analysis and synthesis
- 10 Summary

