

# **EL2520 Control Theory and Practice**

Lecture 5: Multivariable systems

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### From now and on: MIMO

Linear systems with multiple inputs and multiple outputs

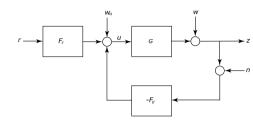
- · Basic properties of multivariable systems
- State-space theory, state feedback and observers
- Decentralized and decoupled control
- Robust loop shaping
- H<sub>2</sub> and H<sub>1</sub> optimal control

The final part of the course considers systems with constraints

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### So far...



#### SISO control revisited:

- Signal norms, system gains and the small gain theorem
- The closed-loop system and the design problem
  - Characterized by six transfer functions: need to look at all!
  - Fundamental limitations and waterbed effect.
- Loop shaping to satisfy sensitivity function specifications.

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## Today's lecture

Basic properties of multivariable systems

- Transfer matrices
- Block diagram calculations
- Gains and directions
- The multivariable frequency response
- Poles and zeros

Chapters 2-3 in the textbook.

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### Transfer matrices

The **Laplace transform** X(s) of a signal x(t) is defined by

$$X(s) = \int_{t=0}^{\infty} x(t)e^{-st} dt$$

Given a linear system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

$$y(t) = Cx(t) + Du(t)$$

and assuming u(t)=0 for t<0 and x(0)=0,

$$Y(s) = \{C(sI - A)^{-1}B + D\}U(s) = G(s)U(s)$$

If system has multiple inputs and outputs, Y and U are vector-valued and **G(s)** is a matrix (i.e. a matrix-valued function of s).

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### Quiz: transfer matrices

1. What is the transfer matrix G(s) for the system

$$\dot{x}_1 = -x_1 + u_1$$

$$\dot{x}_2 = -x_2 + u_2$$

$$y_1 = x_1$$

$$y_2 = x_2$$

How does G(s) change when

2. Input two also affects the first state:  $\dot{x}_1 = -x_1 + u_1 + u_2$ 

3. The second state also affects output one:  $y_1 = x_1 + x_2$ 

4. The second state influences the first:  $\dot{x}_1 = -x_1 - x_2 + u_1$ 

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### Answer and observations

1. 
$$G(s) = \begin{pmatrix} \frac{1}{s+1} & 0\\ 0 & \frac{1}{s+1} \end{pmatrix}$$

Independent subsystems → (block)diagonal transfer matrix

2., 3. 
$$G(s) = \begin{pmatrix} \frac{1}{s+1} & \frac{1}{s+1} \\ 0 & \frac{1}{s+1} \end{pmatrix}$$

Couplings  $\rightarrow$  nondiagonal G(s). Different A, B, C can give same G(s)

4. 
$$G(s) = \begin{pmatrix} \frac{1}{s+1} & -\frac{1}{(s+1)^2} \\ 0 & \frac{1}{s+1} \end{pmatrix}$$

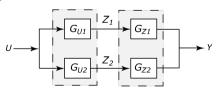
1.-4. have the same poles (eigenvalues of A). Hard to see from G(s)

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### Example: series connection

Linear system viewed as interconnected multivariable systems



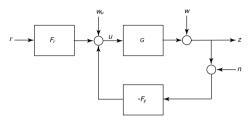
$$\begin{pmatrix} Z_1(s) \\ Z_2(s) \end{pmatrix} = \begin{pmatrix} G_{U1}(s) \\ G_{U2}(s) \end{pmatrix} U(s), \qquad Y(s) = \begin{pmatrix} G_{Z1}(s) & G_{Z2}(s) \end{pmatrix}$$

We see that  $Y(s)=G_Z(s)Z(s)=G_Z(s)G_U(s)U(s)$ .

Note that  $G_ZG_U \leq G_UG_Z$  – **order matters**!

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## Block diagram calculations



Have to be careful when manipulating block diagrams.

**Example.** Let  $w_{ij}$ , w, n=0 and derive transfer matrix from r to z

 $z = Gu = G(F_r r - F_y z) \Rightarrow z + GF_y z = GF_r r$ 

$$z = (I + GF_y)^{-1}GF_rr$$
  $(\neq (I + F_yG)^{-1}GF_r!)$ 

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# Today's lecture

Basic properties of multivariable systems

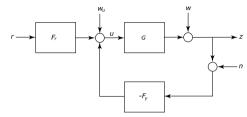
- Transfer matrices
- Block diagram calculations
- · Gains and directions
- The multivariable frequency response
- Poles and zeros

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### Quiz: the closed-loop MIMO system

Determine the sensitivity and complementary sensitivity for the linear multivariable system  $% \left( 1\right) =\left( 1\right) \left( 1\right) \left($ 



**Recall:** S is transfer matrix from  $w \rightarrow z$ , T is transfer matrix from  $-n \rightarrow z$ 

Additional question: What is the relation between S and T?

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### Frequency response and system gain

For a scalar linear system G(s) driven by  $u(t)=\sin(\omega t)$ ,

$$y(t) = |G(i\omega)| \sin(\omega t + \arg G(i\omega))$$

(after transients have died out). So

$$\frac{|Y(i\omega)|_2}{|U(i\omega)|_2} = |G(i\omega)|$$

The system gain (cf. Lecture 1) is defined as

$$\sup_{u} \frac{\|y\|_{2}}{\|u\|_{2}} = \sup_{\omega} |G(i\omega)| = \|G\|_{\infty}$$

Attained for sinusoidal input with frequency  $\omega$  such that  $|G(i\omega)| = ||G||_{\infty}$ 

Q: What are the corresponding results for multivariable systems?

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### Operator norm of linear mapping

Consider the linear mapping y = Ax (x, y, A complex-valued)

Since

$$|y|^2 = |Ax|^2 = (Ax)^*Ax = x^*A^*Ax$$

we have

$$|x|^2 \lambda_{\min}(A^*A) \le |y|^2 \le |x|^2 \lambda_{\max}(A^*A)$$

So

$$\underline{\sigma}(A) \le \frac{|y|}{|x|} \le \overline{\sigma}(A)$$

Where  $\sigma(A) = \sqrt{\lambda(A^*A)}$  are called the **singular values** of A.

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### The singular value decomposition

A matrix A 2  $C^{mf}$  (with r<m, rank(A)=r), can be represented by its singular value decomposition (SVD)

$$A = U\Sigma V^* = \begin{bmatrix} u_1 & \cdots & u_r \end{bmatrix} \operatorname{diag}(\sigma_i) \begin{bmatrix} v_1 & \cdots & v_r \end{bmatrix}^* = \sum_{i=1}^r \sigma_i u_i v_i^*$$

#### where

- The positive scalars  $\sigma_i$  are the **singular values** of A
- v<sub>i</sub> are the **input singular vectors** of A, V\*V=I
- u<sub>i</sub> are the **output singular vectors** of A, U\*U=I

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### SVD interpretation

$$A = U\Sigma V^* = \sum_{i=1}^r \sigma_i u_i v_i^*$$

$$x$$
  $V^*$   $Y^*$   $\Sigma$   $\Sigma$   $V^*$   $U$   $Ax$ 

Interpretation: linear mapping y=Ax can be decomposed as

- compute coefficients of x along input directions v<sub>i</sub>
- scale coefficients by  $\sigma_i$
- reconstitute along output directions u.

Since  $\sigma_1 \leq \cdots \leq \sigma_r$  an input in the  $\mathbf{v}_r$  direction is amplified the most. It generates an output in the direction of  $\mathbf{u}_r$  (typically different from  $\mathbf{v}_r$ )

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### Example

$$G(0) = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0.85 & 0.53 \\ -0.53 & 0.85 \end{bmatrix} \begin{bmatrix} 1.62 & 0 \\ 0 & 0.62 \end{bmatrix} \begin{bmatrix} 0.53 & 0.85 \\ -0.85 & 0.53 \end{bmatrix}^*$$

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# The multivariable frequency response

For a linear multivariable system Y(s)=G(s)U(s), we have

$$Y(i\omega) = G(i\omega)U(i\omega)$$

Since this is a linear mapping,

$$\underline{\sigma}(G(i\omega)) \le \frac{|Y(i\omega)|}{|U(i\omega)|} \le \overline{\sigma}(G(i\omega))$$

with equality if  $U(i\omega)$  parallell w. corresponding input singular vector.

For example,

$$\frac{|Y(i\omega)|}{|U(i\omega)|} = \overline{\sigma}(G(i\omega))$$

only if  $U(i\omega)$  parallell with input singular vector corresponding to  $\overline{\sigma}$ 

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### The system gain

As for scalar systems, we can use Parseval's theorem to find

$$||y||_2 \le ||G||_\infty ||u||_2$$

where

$$||G||_{\infty} = \sup_{\omega} |G(i\omega)| = \sup_{\omega} \overline{\sigma}(G(i\omega))$$

**Note:** Worst-case input is sinusoidal at the frequency that attains the supremum, but its components are appropriately scaled and phase shifted (as specified by the input singular vector of  $\overline{\sigma}$ )

**Note:** the infinity norm computes the maximum amplifications across frequency ( $\sup_{m}$ ) and input directions ( $\overline{\sigma}$ )

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### Example: heat exchanger

**Objective:** control outlet temperatures  $T_{c\prime}$   $T_{H}$  by manipulating the flows  $q_{C\prime}$   $q_{H}$ 



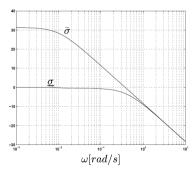
Model:

$$\begin{pmatrix} T_C \\ T_H \end{pmatrix} = \frac{1}{(100s+1)(2.5s+1)} \begin{pmatrix} -19(5s+1) & 18 \\ -18 & 19(5s+1) \end{pmatrix} \begin{pmatrix} q_C \\ q_H \end{pmatrix}$$

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## Singular values of heat exchanger



System gain:  $||G||_{\infty}=31 \text{ dB } (=37)$ 

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### Heat exchanger steady-state

$$G(0) = \begin{bmatrix} -19 & 18 \\ -18 & 19 \end{bmatrix} = \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 37 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$$

Singular value decomposition reveals

• Maximum effect input (input singular vector corresponding to  $\overline{\sigma}$  )

$$\begin{bmatrix} q_C \\ q_H \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow \begin{bmatrix} T_C \\ T_H \end{bmatrix} = \begin{bmatrix} 37 \\ 37 \end{bmatrix}$$

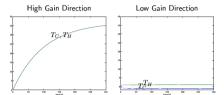
• Minimum effect input (corresponding to  $\sigma$ )

$$\begin{bmatrix} q_C \\ q_H \end{bmatrix} = \begin{bmatrix} \mathbf{1} \\ \mathbf{1} \end{bmatrix} \Rightarrow \begin{bmatrix} T_C \\ T_H \end{bmatrix} = \begin{bmatrix} -\mathbf{1} \\ \mathbf{1} \end{bmatrix}$$

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### Heat exchanger step responses



Input direction has dramatic effect! (agrees with physical intuition)

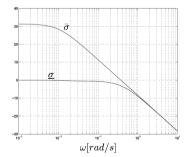
Note: large difference in time-scales!

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## Singular values and bandwidths

What is the bandwidth of the system?



No single value, but a range. Depends on input directions.

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## Today's lecture

### Basic properties of multivariable systems

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### Poles

**Definition.** The **poles** of a linear systems are the eigenvalues of the system matrix in a minimal state-space realization.

**Definition.** The **pole polynomial** is the characteristic polynomial of the A matrix,  $\lambda(s) = \det(sI-A)$ .

Alternatively, the poles of a linear system are the zeros of the pole polynomial, i.e., the values  $\mathbf{p}_i$  such that  $\lambda(p_i)=0$ 

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### Poles of multivariable systems

**Theorem.** The pole polynomial of a system with transfer matrix G(s) is the common denominator of all minors of G(s)

**Recall:** a minor of a matrix M is the determinant of a (smaller) square matrix obtained by deleting some rows and columns of M

**Example:** The minors of

$$G(s) = \begin{bmatrix} \frac{2}{s+1} & \frac{3}{s+2} \\ \frac{1}{s+1} & \frac{1}{s+1} \end{bmatrix}$$

are 
$$\frac{2}{s+1}$$
,  $\frac{3}{s+2}$ ,  $\frac{1}{s+1}$  and  $\det G(s) = \frac{1-s}{(s+1)^2(s+2)}$ 

Thus, the system has poles in s=-1 (a double pole) and s=-2.

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### Poles cont'd

Since the transfer matrix is given by

$$G(s) = C(sI - A)^{-1}B + D = \frac{1}{\det(sI - A)}r(s)$$

where r(s) is a polynomial in s (see book for precise expression), the pole polynomial must be "at least" the least common denominator of the the elements of the transfer matrix.

**Example:** The system

$$G(s) = \begin{bmatrix} \frac{2}{s+1} & \frac{3}{s+2} \\ \frac{1}{s+1} & \frac{1}{s+1} \end{bmatrix} = \frac{1}{(s+1)(s+2)} \begin{bmatrix} 2(s+2) & 3(s+1) \\ (s+2) & (s+2) \end{bmatrix}$$

must (at least) have poles in s=-1 and s=-2.

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### Zeros

**Theorem.** The **zero polynomial** of G(s) is the greatest common divisor of the maximal minors of G(s), normed so that they have the pole polynomial of G(s) as denominator. The **zeros** of G(s) are the roots of its zero polynomial.

Example: The maximal minor of

$$G(s) = \begin{bmatrix} \frac{2}{s+1} & \frac{3}{s+2} \\ \frac{1}{s+1} & \frac{1}{s+1} \end{bmatrix}$$

is  $\det G(s) = \frac{1-s}{(s+1)^2(s+2)}$  (already normed!).

Thus, G(s) has a zero at s=1.

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## Quiz: multivariable poles and zeros

What are the poles and zeros of the multivariable system

$$G(s) = \frac{1}{(s+1)} \begin{pmatrix} 1 & s+1 \\ s-1 & 1 \end{pmatrix}$$

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### **Summary**

An introduction to multivariable linear systems:

- Block diagram manipulations (order matters!)
- System gain (directions matter!)
- Poles and zeros

More next week!

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### Notes on poles and zeros

For scalar system G(s) with poles  $p_i$  and zeros  $z_i$ ,

$$G(z_i) = 0, \quad G(p_i) = \infty$$

For a multivariable system, directions matter!

For a system with pole p, there exist vectors  $u_p$ ,  $v_p$ :

$$u_p^*G(p) = \infty$$
  $G(p)v_p = \infty$ 

Similarly, a zero at  $z_i$  implies the existence of vectors  $u_z$ ,  $v_z$ :

$$u_z^*G(z) = 0 \qquad G(z)v_z = 0$$

As for scalar systems, a zero at s=z implies that there exists a signal on the form  $u(t)=v_ze^{-zt}$  for t, 0, and u(t)=0 for t<0, and initial values  $x(0)=x_z$  so that y(t)=0 for t, 0

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