



# DISTANCE BETWEEN LINEAR SYSTEMS

and

# ORDER REDUCTION

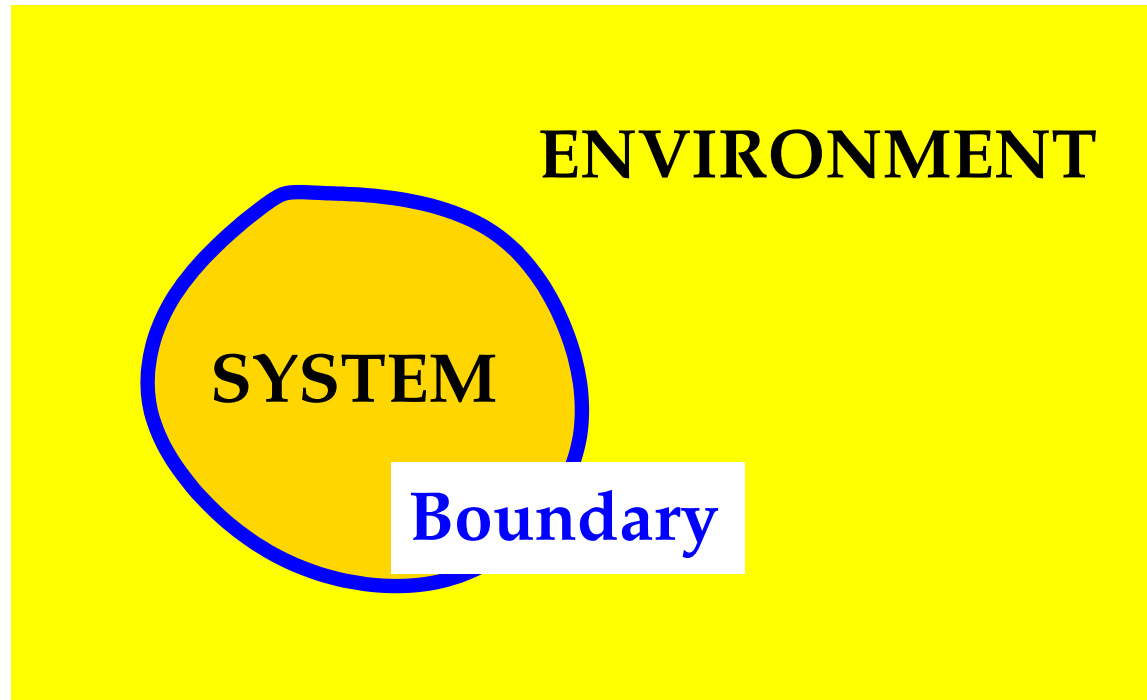
**Jan C. Willems**  
**ESAT, K.U. Leuven, Flanders, Belgium**

In honor of Adhemar Bultheel on the occasion of his 60th birthday



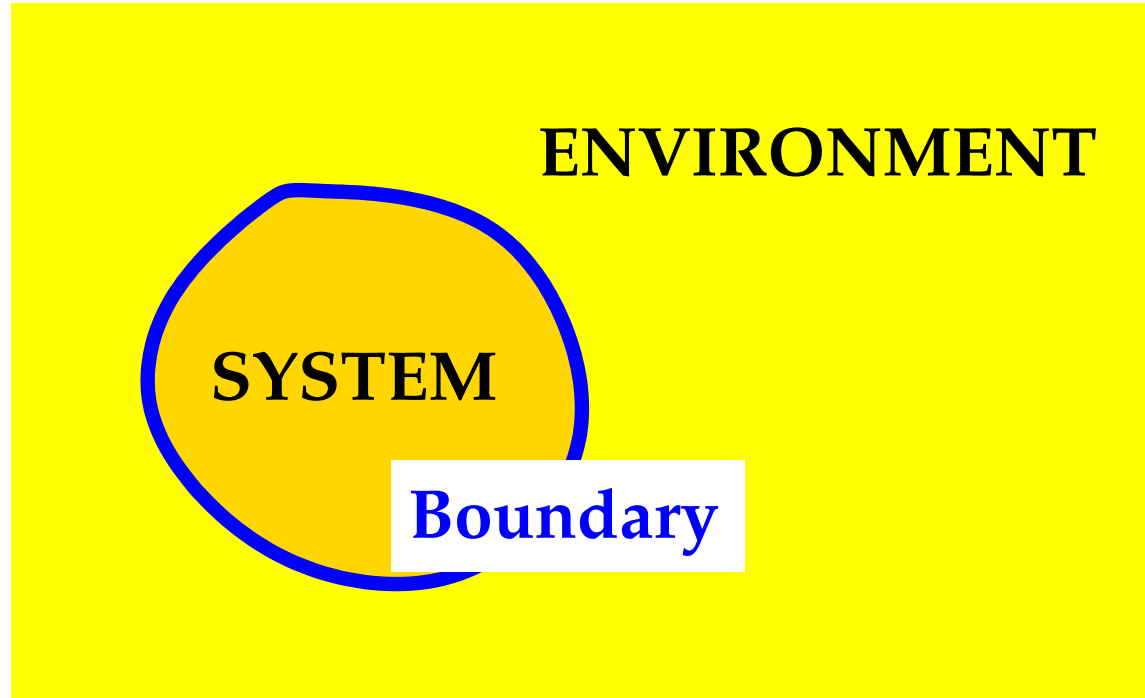
# Open systems

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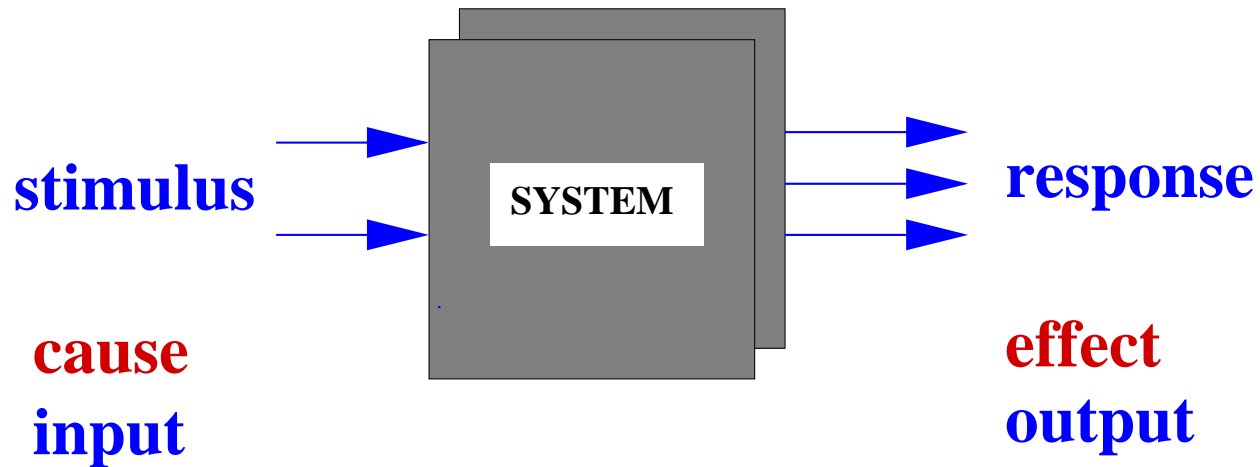
**Systems interact with their environment**

# Open systems



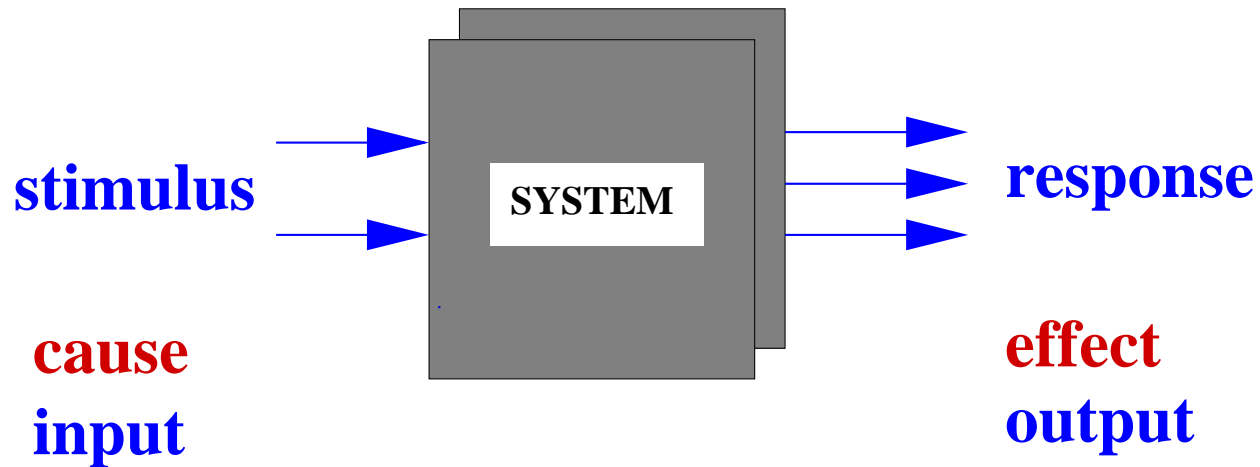
**How are open systems formalized ?**

# Classical approach: input/output systems

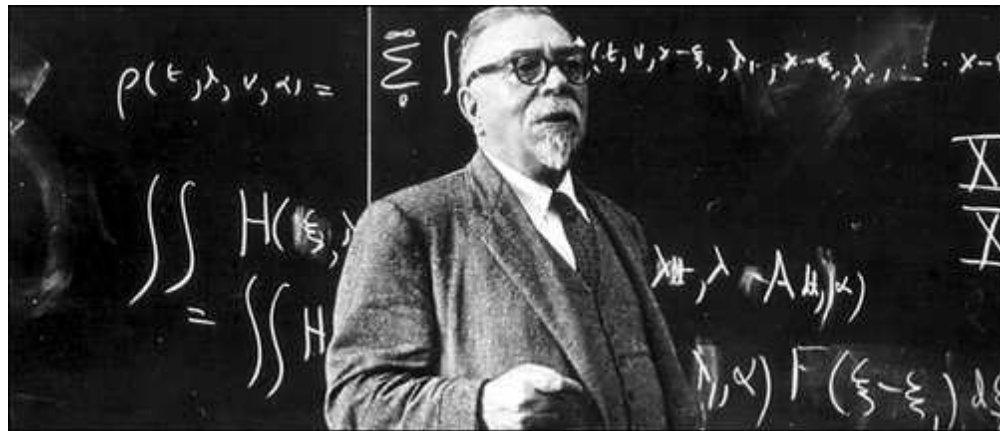


**Convolutions, transfer functions, impedances, ...**

# Classical approach: input/output systems



**Oliver Heaviside**  
(1850-1925)



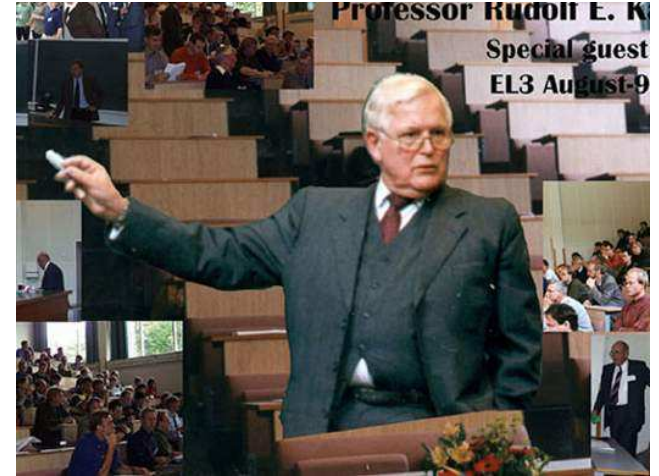
**Norbert Wiener (1894-1964)**

**and many electrical circuit theorists**

# Input/state/output systems

Around 1960: a **paradigm shift** to

$$\frac{d}{dt}\mathbf{x} = f(\mathbf{x}, u), \quad y = g(\mathbf{x}, u)$$



Rudolf Kalman (1930- )

**This framework turned out to be very effective and useful, for example in model order reduction (MOR).**



## Model order reduction

**MOR for linear systems ( $\cong$  rational approximation)**

$$\frac{d}{dt}x = Ax + Bu, y = Cx, \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m, y \in \mathbb{R}^p$$

**Assume stable (A Hurwitz: roots in open left half of  $\mathbb{C}$ ).**

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$$u \mapsto y$$

$$y(t) = \int_{-\infty}^t C e^{A(t-t')} B u(t') dt'$$

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**∴ Approximate this system, this map, by another one**

$$u \mapsto y \quad y(t) = \int_{-\infty}^t C_{\text{red}} e^{A_{\text{red}}(t-t')} B_{\text{red}} u(t') dt'$$

**‘simpler’: lower state dim. !!**  $x_{\text{red}} \in \mathbb{R}^{n_{\text{red}}}$ ,  $n_{\text{red}} \ll n$   
 $\rightsquigarrow$  **balancing, AAK, Krylov, POD, etc.**

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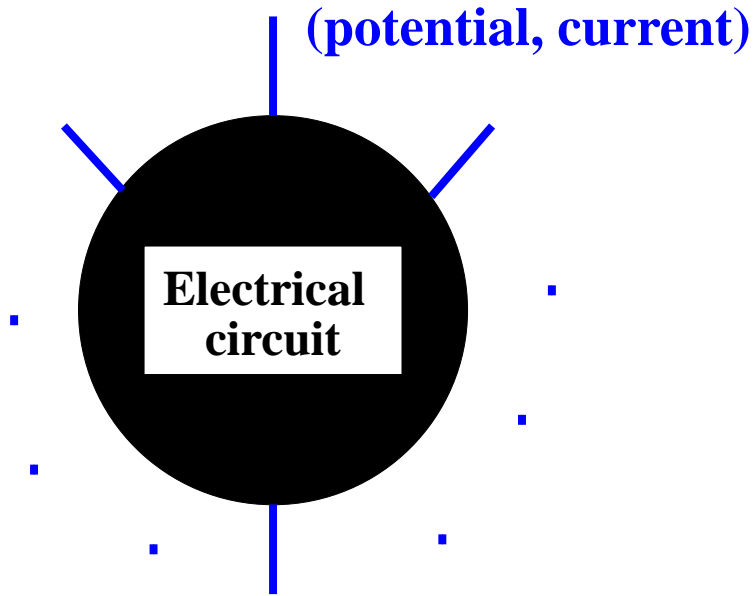
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**$\exists$  effective methods for MOR for  $\boxed{\text{stable}}$  LTI i/o systems**

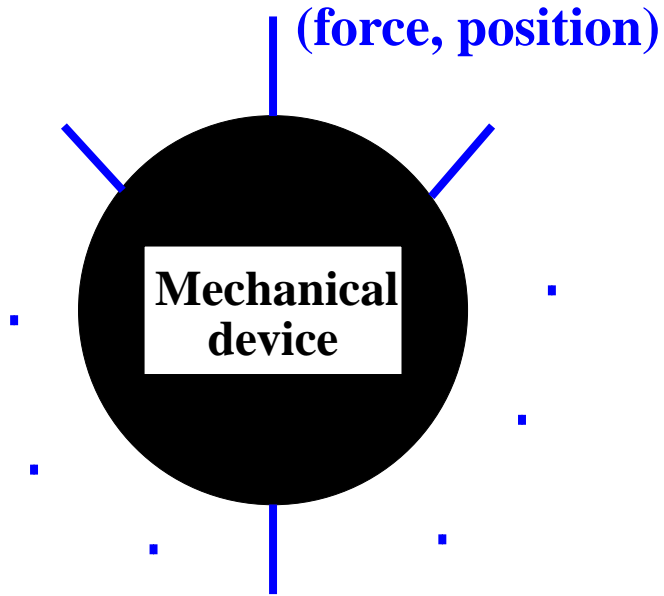
# **Drawbacks of input/output thinking**

## Physical systems with terminals



Associated with each terminal  
there are **two** variables.  
Which should be considered  
**input? output?**

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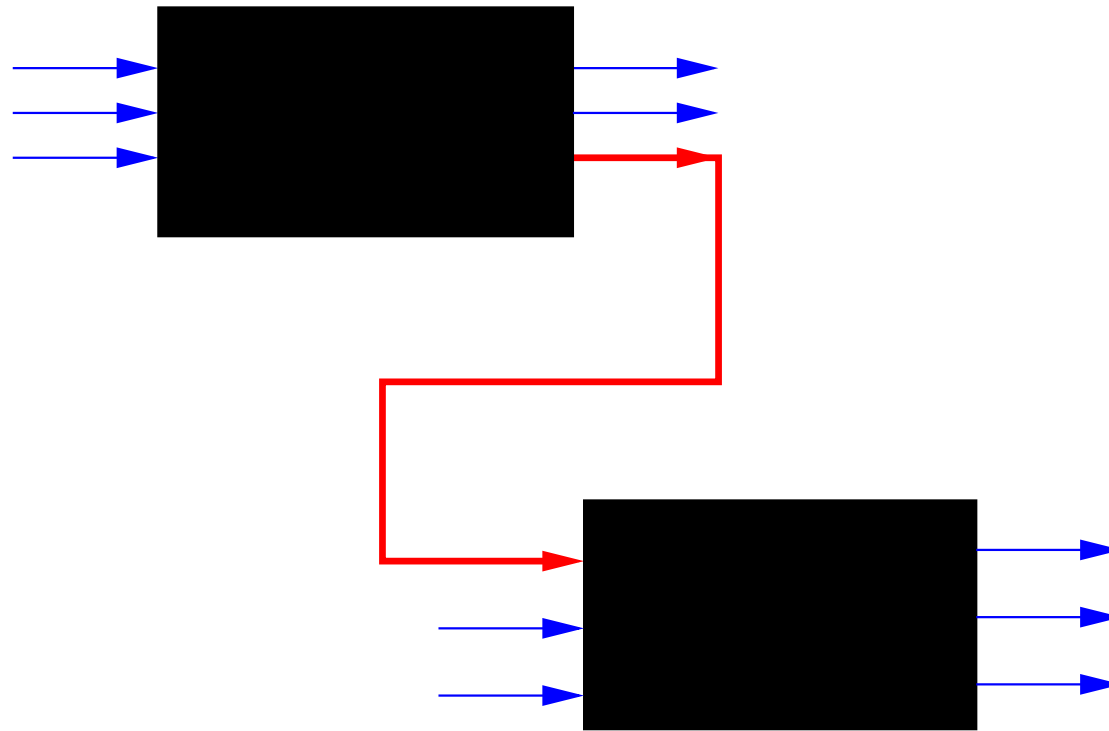


Associated with each terminal  
there are **two** variables.  
Which should be considered  
**input? output?**

- ▶ mechanical systems (terminal var's: force & position)
- ▶ *et cetera*

# System interconnection as output-to-input assignment

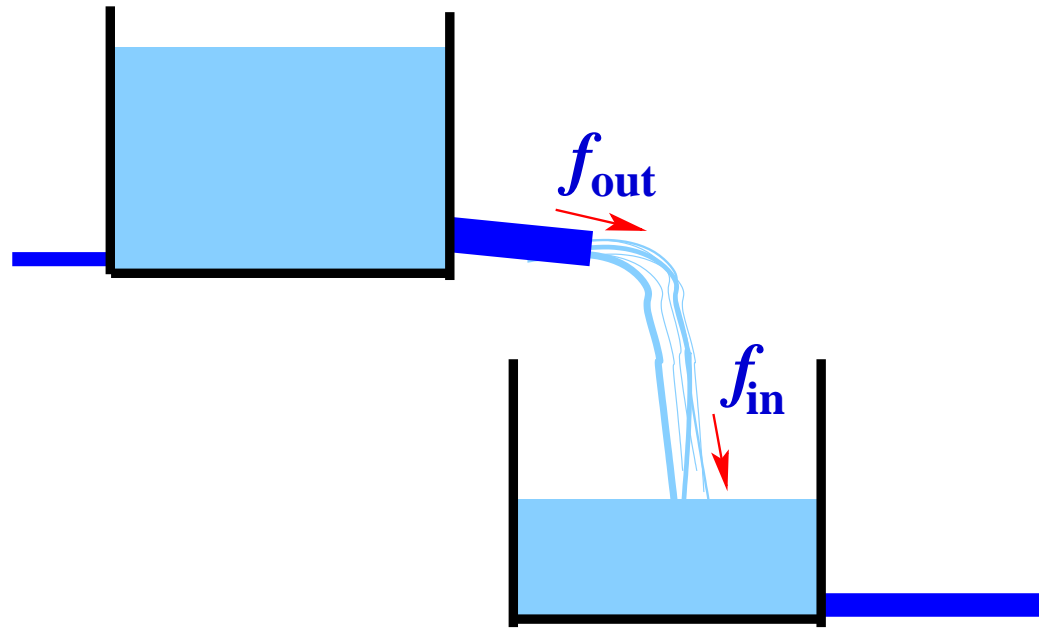
**The classical view of system interconnection:**





# System interconnection as output-to-input assignment

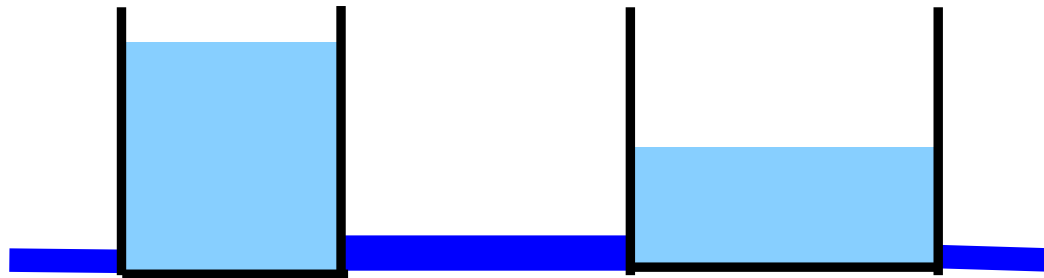
There are **many** examples where output-to-input connection is eminently natural.



## System interconnection as output-to-input assignment

There are **many** examples where output-to-input connection is eminently natural.

But for other interconnections, i/o is more problematic.



**Interconnection = variable sharing, not output-to-input assignment**

## Ceterum censeo

**The input/output approach as the primary and universal view of open systems is a misconception.**

**Physical systems are not signal processors !**

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**Physical systems are not signal processors !**

**How should we formalize open systems, if not as input/output systems?**

**How does MOR function then?**

# Linear time-invariant differential systems

**LTIDSs**

## LTIDSs

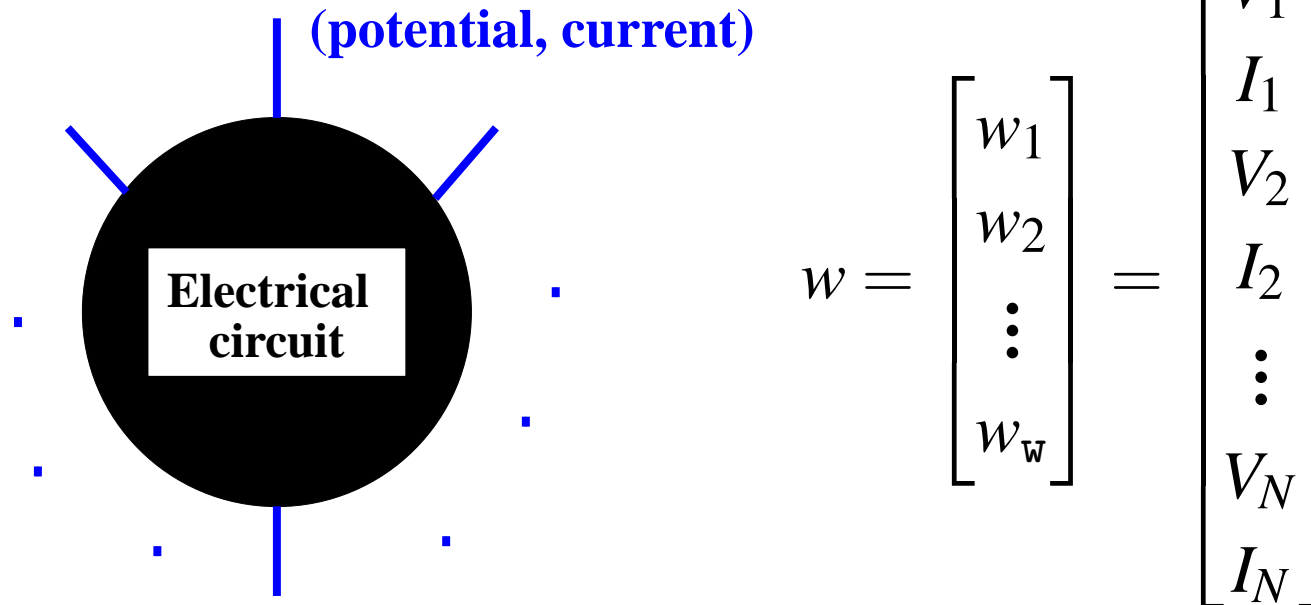
**We consider systems described by linear, constant-coefficient, differential equations**

$$R \left( \frac{d}{dt} \right) w = 0$$

**with  $R$  a polynomial matrix,  $R \in \mathbb{R}[\xi]^{\bullet \times w}$**

We consider systems described by linear, constant-coefficient, differential equations

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All system variables are treated on the same footing.

A model = a relation (rather than a map)

## LTIDSs

We consider systems described by linear, constant-coefficient, differential equations

$$R \left( \frac{d}{dt} \right) w = 0$$

**Behavior**  $\mathcal{B} :=$  the set of solutions

If you so like, assume the solutions in  $\mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w)$ .



## LTIDSs - Rational symbol representation

$$R \left( \frac{d}{dt} \right) w = 0$$

readily generalized to the case where  $R$  is a matrix of  
**rational functions**  $R \in \mathbb{R}(\xi)^{\bullet \times \mathbf{w}}$

$$R = P^{-1}Q, \quad P, Q \in \mathbb{R}[\xi]^{\bullet \times \bullet}, \quad \text{left coprime}$$

$$R \left( \frac{d}{dt} \right) w = 0 \quad :\Leftrightarrow \quad Q \left( \frac{d}{dt} \right) w = 0$$

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$\rightsquigarrow$  **A LTID behavior has many representations.**

**A very special representation**

## Norm-preserving image representations

**Let  $\mathcal{B}$  be the behavior of a controllable LTIDS.  
Then it allows an observable ‘image representation’**

$$w = M\left(\frac{d}{dt}\right)\ell \quad \text{with } M \in \mathbb{R}(\xi)^{w \times \bullet}$$

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**norm-preserving image representation**

**i.e.** 
$$\int_{-\infty}^{+\infty} \|w(t)\|^2 dt = \int_{-\infty}^{+\infty} \|\ell(t)\|^2 dt$$

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norm-preserving image representation

$$\text{i.e.} \quad \int_{-\infty}^{+\infty} \|w(t)\|^2 dt = \int_{-\infty}^{+\infty} \|\ell(t)\|^2 dt$$

$M$  cannot be polynomial, **it must be rational**.

Obviously  $M$  must also be proper.

Can also make  $M$  **stable** (meaning: its poles are in the left half of the complex plane).

# **Distance between behaviors**

## Distance between subspaces

**A model is a behavior, a set (of trajectories).**

**Hence the distance between LTIDSs translates into the distance between **linear subspaces**.**

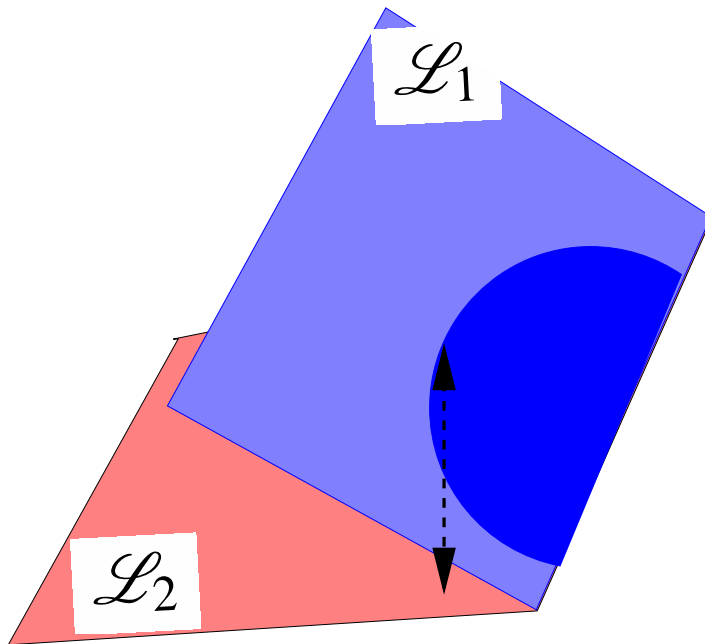


## Distance between subspaces

Hence the distance between LTIDSs translates into the distance between **linear subspaces**.

$\mathcal{L}_1, \mathcal{L}_2$  linear subspaces of a Hilbert space.

$$\overrightarrow{d}(\mathcal{L}_1, \mathcal{L}_2) := \sup_{x_1 \in \mathcal{L}_1, \|x_1\|=1} \inf_{x_2 \in \mathcal{L}_2} \|x_1 - x_2\|$$



**closest point**  
on unit sphere of  $\mathcal{L}_1$   
from  $\mathcal{L}_2$

## Distance between subspaces

$$\mathbf{gap}(\mathcal{L}_1, \mathcal{L}_2) := \mathbf{max} \left\{ \overrightarrow{d}(\mathcal{L}_1, \mathcal{L}_2), \overrightarrow{d}(\mathcal{L}_2, \mathcal{L}_1) \right\}$$

$$0 \leq \mathbf{gap}(\mathcal{L}_1, \mathcal{L}_2) \leq 1$$

## Association of a subspace of a Hilbert space to a LTID behavior

**The behavior  $\mathcal{B}$  of a LTIDS is not a subspace of a Hilbert space.**

**Which subspace of which Hilbert space should we associate with a LTID behavior  $\mathcal{B}$ ?**

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**Which subspace of which Hilbert space should we associate with a LTID behavior  $\mathcal{B}$ ?**

$$\mathcal{B} \mapsto \mathcal{B}^{\mathcal{L}_2} := \mathcal{B} \cap \mathcal{L}_2(\mathbb{R}, \mathbb{R}^w)$$

## Distance between behaviors

Define the distance between two LTID behaviors as

$$d(\mathcal{B}_1, \mathcal{B}_2) := \text{gap}(\mathcal{B}_1^{\mathcal{L}_2}, \mathcal{B}_2^{\mathcal{L}_2})$$

So, we consider the  $\mathcal{L}_2$ -trajectories for measuring distance.

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Keep notation  $\mathcal{B}$  for  $\mathcal{B}^{\mathcal{L}_2} = \mathcal{B} \cap \mathcal{L}_2(\mathbb{R}, \mathbb{R}^w)$ .

$$\forall w_1 \in \mathcal{B}_1, \exists w_2 \in \mathcal{B}_2 \text{ such that } \|w_1 - w_2\| \leq \mathbf{gap}(\mathcal{B}_1, \mathcal{B}_2) \|w_1\|$$

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Small gap  $\Rightarrow$  the LTIDSs are ‘close’.

‘Phenomena’ in  $\mathcal{B}_1$  are well approximated by ‘phenomena’ in  $\mathcal{B}_2$ , and vice-versa.

# Computation of the gap



## Distance between LTID behaviors

- ▶ **How to compute the gap?**
- ▶ **Model reduce according to the gap!**

## Formula for the gap for LTID behaviors

$\mathcal{B}_1, \mathcal{B}_2$       **LTID behaviors.**

**Take norm-preserving image representations**

$$w = M_1\left(\frac{d}{dt}\right)\ell_1, \quad w = M_2\left(\frac{d}{dt}\right)\ell_2$$

**Then**

$$\begin{aligned} \mathbf{gap}(\mathcal{B}_1, \mathcal{B}_2) &= \left\| M_1(i\omega)M_1(-i\omega)^\top - M_2(i\omega)M_2(-i\omega)^\top \right\|_{\mathcal{L}_\infty} \\ &\leq \left\| M_1(i\omega) - M_2(i\omega) \right\|_{\mathcal{L}_\infty} \end{aligned}$$

**MOR in the gap**

## Balanced MOR

Let  $M$  be the transfer function of a stable input/output system (strictly proper rational function,  $n$  poles, all in open left half plane).

Associated with  $M$  there are nonnegative real numbers (the **Hankel singular values**)

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_{n_{\text{red}}} \geq \dots \geq \sigma_n > 0$$

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$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_{n_{\text{red}}} \geq \dots \geq \sigma_n > 0$$

leading to  $M_{\text{red}}$  stable input/output reduced system with  $n_{\text{red}}$  poles, in LHP.

Balanced model reduction  $\Rightarrow$

$$\|M(i\omega) - M_{\text{red}}(i\omega)\|_{\mathcal{H}_\infty} \leq 2 \sum_{\text{neglected Hankel SVs of } G} \sigma_k$$

## Reduction of a stable norm-preserving representation

Start with a LTID behavior  $\mathcal{B}$ . Represent  $\mathcal{B}$  by a **norm-preserving, stable** image representation

$$\boxed{w = M\left(\frac{d}{dt}\right)\ell} \quad \text{with } M \in \mathbb{R}(\xi)^{w \times \bullet}$$

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Now MOR (in the sense of the state dimension  $\cong$  the order of the underlying ODE), in the classical way, viewed as a stable input/output system (input  $\ell$ , output  $w$ ) using balancing

$$\rightsquigarrow \boxed{w = M_{\text{red}}\left(\frac{d}{dt}\right)\ell}$$

Error bound (classical - **‘twice the sum of the tail’**):

$$\|M(i\omega) - M_{\text{red}}(i\omega)\|_{\mathcal{H}_\infty} \leq 2 \sum_{\text{neglected Hankel SVs of } M} \sigma_k$$

## Behavior approximation and gap error bound

Start with stable norm-preserving representation of  $\mathcal{B}$

$$w = M\left(\frac{d}{dt}\right)\ell \quad \text{with } M \in \mathbb{R}(\xi)^{w \times \bullet}$$

MOR using balancing  $\rightsquigarrow w = M_{\text{red}}\left(\frac{d}{dt}\right)\ell$ .

Call the behavior of the reduced system  $\mathcal{B}_{\text{red}}$ .



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Error bound

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## Gap error bound

$\forall w \in \mathcal{B} \exists w' \in \mathcal{B}_{\text{red}}$  such that

$$\|w - w'\| \leq \left( 2 \sum_{\text{neglected Hankel SVs of } M} \sigma_k \right) \|w\|$$

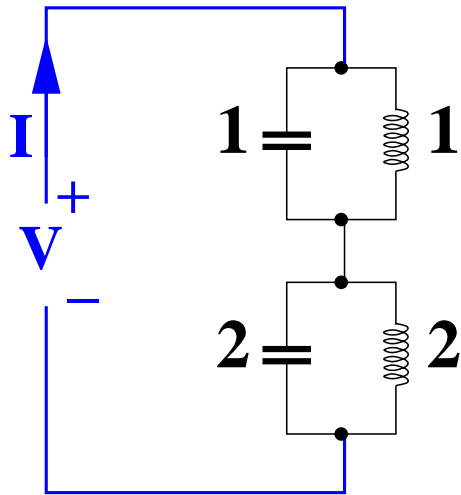
and vice-versa.

$\sum_{\text{neglected Hankel SVs of } M} \sigma_k$  small

$\Rightarrow$  as linear subspaces,  
 $\mathcal{B}_{\text{red}}$  is a good approximation of  $\mathcal{B}$   
in the gap metric.

**Example**

# LCLC circuit

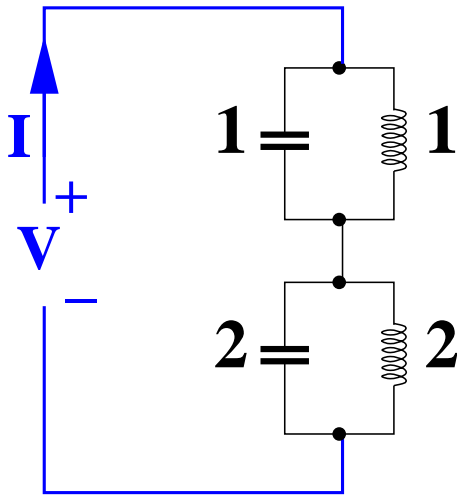


**system order = 4. Reduce to 2!**

**kernel**  $\left(1 + 5\frac{d^2}{dt^2} + 4\frac{d^4}{dt^4}\right) V = \left(3\frac{d}{dt} + 6\frac{d^3}{dt^3}\right) I$

**image**  $\begin{bmatrix} I \\ V \end{bmatrix} = \begin{bmatrix} 1 + 5\frac{d^2}{dt^2} + 4\frac{d^4}{dt^4} \\ 3\frac{d}{dt} + 6\frac{d^3}{dt^3} \end{bmatrix} \ell$

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**stable norm-preserving image**

$$\begin{bmatrix} I \\ V \end{bmatrix} = \frac{1}{1 + 3\frac{d}{dt} + 5\frac{d^2}{dt^2} + 6\frac{d^3}{dt^3} + 4\frac{d^4}{dt^4}} \begin{bmatrix} 1 + 5\frac{d^2}{dt^2} + 4\frac{d^4}{dt^4} \\ 3\frac{d}{dt} + 6\frac{d^3}{dt^3} \end{bmatrix} \ell$$

**Apply balancing algorithm**  $\rightsquigarrow$

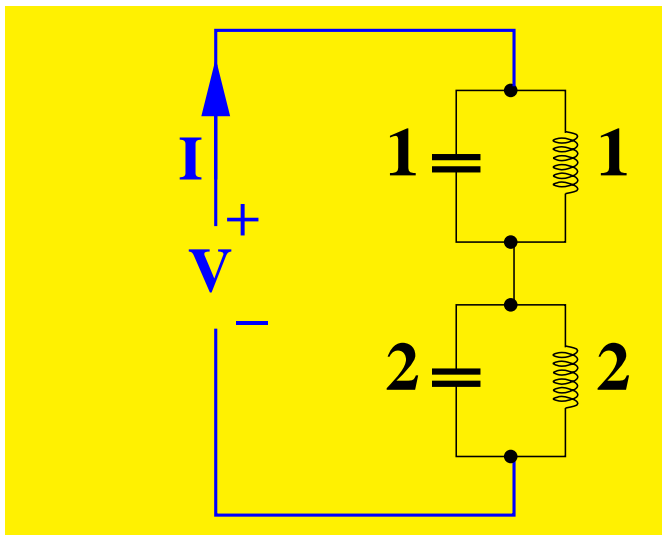
# LCLC circuit

stable norm-preserving image

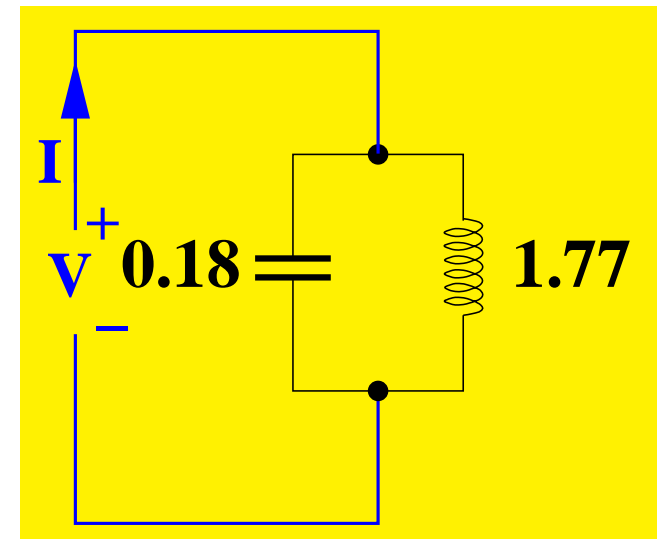
$$\begin{bmatrix} I \\ V \end{bmatrix} = \frac{1}{1 + 3\frac{d}{dt} + 5\frac{d^2}{dt^2} + 6\frac{d^3}{dt^3} + 4\frac{d^4}{dt^4}} \begin{bmatrix} 1 + 5\frac{d^2}{dt^2} + 4\frac{d^4}{dt^4} \\ 3\frac{d}{dt} + 6\frac{d^3}{dt^3} \end{bmatrix} \ell$$

red. order = 2

$$\begin{bmatrix} I \\ V \end{bmatrix} = \frac{1}{\frac{d^2}{dt^2} + 0.1861\frac{d}{dt} + 0.3298} \begin{bmatrix} \frac{d^2}{dt^2} + 0.3298 \\ 0.1861\frac{d}{dt} \end{bmatrix} \ell$$



~>



## Recapitulation

- ▶ **The gap is a measure of the distance between closed linear subspaces of a Hilbert space.**
- ▶ **Through the  $\mathcal{L}_2$  behavior, the gap gives a measure of the distance between controllable LTIDSs.**

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- ▶ **The gap is a measure of the distance between closed linear subspaces of a Hilbert space.**
- ▶ **Through the  $\mathcal{L}_2$  behavior, the gap gives a measure of the distance between controllable LTIDSs.**
- ▶ **Observable norm-preserving image representations of LTIDSs allow to compute the gap,**
- ▶ **and lead to a model reduction algorithm with an error bound in the gap.**



**The lecture frames are available from/at**

`http://www.esat.kuleuven.be/~jwillems`

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**Thank you**

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**!!! AD MULTOS ANNOS FELICES !!!**



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