



# OPEN DYNAMICAL SYSTEMS: Basic concepts and examples

Chaire Francqui, Lecture I, May 5, 2004

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# THEMES

- **Mathematical models, the behavior**
- **Dynamical systems**
- **Examples**
- **Linear differential systems  $\rightsquigarrow$  3 basic theorems:**
  1. **Characterization as submodules and Structure of kernel representations**
  2. **Elimination theorem**
  3. **Input/output representability**

# Generalities

What is a model? As a **mathematical** concept.

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What is a model?

What is a **dynamical** system? As a **mathematical** concept.

# Generalities

What is a model?

What is a **dynamical** system?

What is the role of **differential equations** in thinking about dynamical models?

# Generalities

## Intuition

We have a ‘phenomenon’ that produces ‘outcomes’ (‘events’).  
We wish to model the outcomes that can occur.

**Before** we model the phenomenon:

the outcomes are in a set, which we call the *universum*.

**After** we model the phenomenon:

the outcomes are declared (thought, believed)  
to belong to the *behavior* of the model,  
a subset of the universum.

**This subset is what we call the mathematical model.**

## Generalities

This way we arrive at the

### Definition

A *mathematical model* is a subset  $\mathcal{B}$  of a universum  $\mathcal{U}$  of outcomes,

$$\mathcal{B} \subseteq \mathcal{U}.$$

$\mathcal{B}$  is called the *behavior* of the model.

# Features

- **Generality, applicability**
- **shows the role of model equations**
- $\rightsquigarrow$  **notion of equivalent models**
- $\rightsquigarrow$  **notion of more powerful model**
- **Structure, symmetries**
- **...**



In these lectures, we will only consider **deterministic** models.

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Stochastic models: there is a map  $P$  (the 'probability')

$$P : \mathcal{A} \rightarrow [0, 1]$$

with  $\mathcal{A}$  a ' $\sigma$ -algebra' of subsets of  $\mathcal{U}$ .

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$P(\mathfrak{B}) =$  'the degree of certainty (belief, plausibility,  
relative frequency) that outcomes are in  $\mathfrak{B}$ ';

$\cong$  the degree of validity of  $\mathfrak{B}$  as a model.

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$\mu(x) =$  'the extent to which  $x \in \mathcal{U}$  belongs to the model'.

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Determinism:  $\mathcal{A} = \{, \mathfrak{B}, \mathfrak{B}^{\text{complement}}, \mathcal{U}\}$ .

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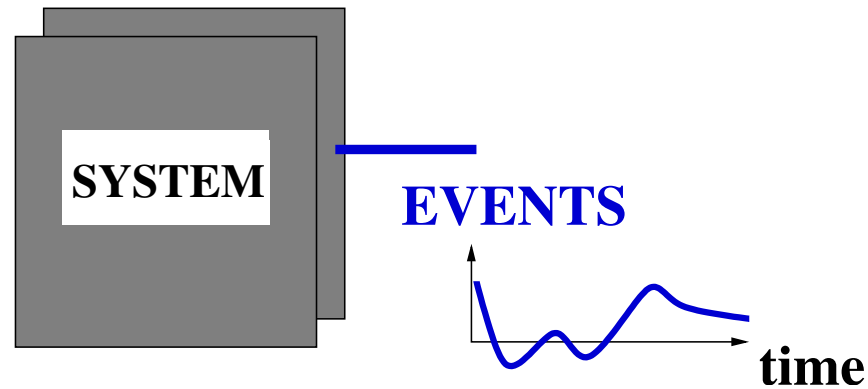
$$\mu : \mathcal{U} \rightarrow [0, 1]$$

Determinism:  $\mu$  is 'crisp':

$$\text{image}(\mu) = \{0, 1\}, \quad \mathfrak{B} = \mu^{-1}(\{1\}) := \{x \in \mathcal{U} \mid \mu(x) = 1\}$$

# Dynamical systems

In dynamics, the outcomes are functions of time  $\rightsquigarrow$



**Which event trajectories are possible?**

# Dynamical systems

## Definition

A dynamical system =  $\Sigma := (\mathbb{T}, \mathbb{W}, \mathcal{B})$

with  $\mathbb{T} \subseteq \mathbb{R}$ , the *time-axis* (= the relevant time instances),  
 $\mathbb{W}$ , the *signal space*

(= where the variables take on their values),

$\mathcal{B} \subseteq \mathbb{W}^{\mathbb{T}}$  the *behavior* (= the admissible trajectories).



# Dynamical systems

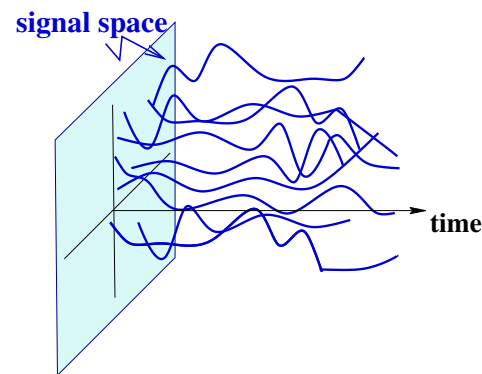
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**Totality of 'legal' trajectories =: the behavior**

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For a trajectory ('an event')  $w : \mathbb{T} \rightarrow \mathbb{W}$ , we thus have:

$w \in \mathcal{B}$  : the model **allows** the trajectory  $w$ ,

$w \notin \mathcal{B}$  : the model **forbids** the trajectory  $w$ .

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Usually,

$\mathbb{T} = \mathbb{R}$ , or  $[0, \infty)$ , etc. (in continuous-time systems),  
or  $\mathbb{Z}$ , or  $\mathbb{N}$ , etc. (in discrete-time systems).

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Usually,

$\mathbb{W} \subseteq \mathbb{R}^w$  (in lumped systems),

a function space

(in distributed systems, time a distinguished variable),

a finite set (in DES)' etc.

# Dynamical systems

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## Emphasis:

$$\mathbb{T} = \mathbb{R},$$

$$\mathbb{W} = \mathbb{R}^w,$$

$\mathcal{B}$  = solutions of system of linear constant coefficient  
ODE's, or difference eqn's, or PDE's.

## More structure

$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

is said to be

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$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

is said to be **linear**

if  $\mathbb{W}$  is a vector space, and  $\mathfrak{B}$  a linear subspace of  $\mathbb{W}^{\mathbb{T}}$ .

## More structure

$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

is said to be **time-invariant**

if  $\mathbb{T} = \mathbb{R}, \mathbb{R}_+, \mathbb{Z}$ , or  $\mathbb{Z}_+$  and if  $\mathfrak{B}$  satisfies

$$\sigma^t \mathfrak{B} \subseteq \mathfrak{B} \text{ for all } t \in \mathbb{T}.$$

$\sigma^t$  denotes the **shift**,  $\sigma^t f(t') := f(t' + t)$ .



## More structure

$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

is said to be **differential**

if  $\mathbb{T} = \mathbb{R}$ , or  $\mathbb{R}_+$ , etc., and if  $\mathfrak{B}$  is the solution set of a (system of) ODE's.

a **difference system** if, etc.

## More structure

$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

is said to be

**symmetric**

w.r.t. the transformation group  $\{T_g, g \in \mathfrak{G}\}$  on  $\mathbb{W}^{\mathbb{T}}$

if  $T_g \mathfrak{B} = \mathfrak{B}$  for all  $g \in \mathfrak{G}$ .

### Examples:

1. time-invariance, time-reversibility
2. permutation symmetry, rotation symmetry, translation symmetry, Euclidean symmetry,
3. etc., etc.

## **A series of examples**

# Planetary motion

**Let's put Kepler and Newton in this setting.**

# Planetary motion

Let's put Kepler and Newton in this setting.



How does it move?

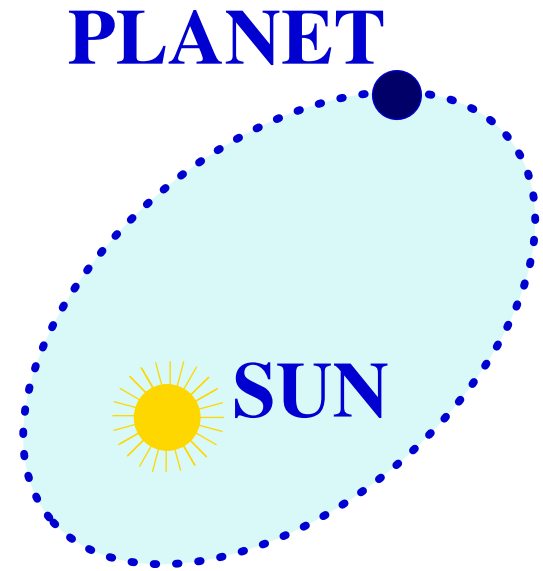
# Kepler's laws



**Johannes Kepler (1571-1630)**

**Kepler's first law:**

**Ellipse, sun in focus**



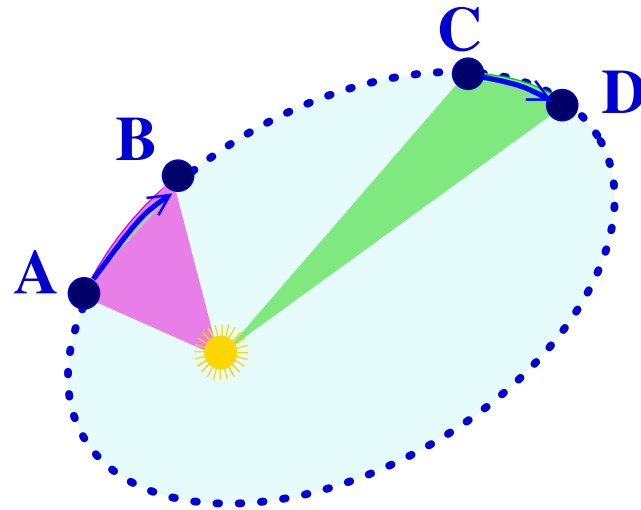
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Kepler's second law:

= areas in = times



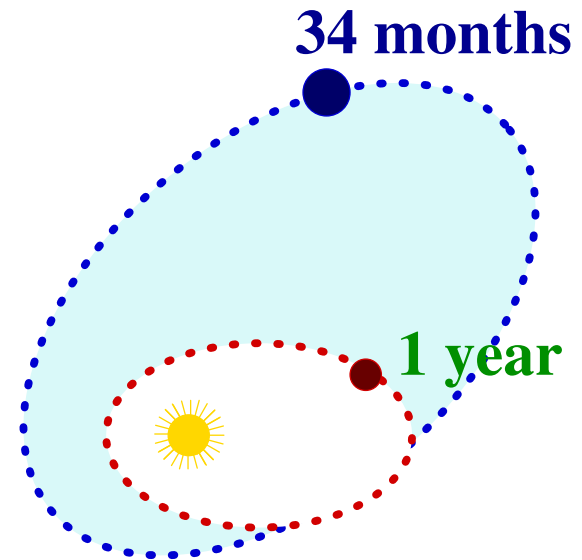
# Kepler's laws



Johannes Kepler (1571-1630)

Kepler's third law:

$$(\text{period})^2 = (\text{diameter})^3$$





## Kepler's laws

This obviously defines a dynamical system  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$$\mathbb{T} = \mathbb{R},$$

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$\mathfrak{B} =$  all  $w : \mathbb{R} \rightarrow \mathbb{R}^3$  that satisfy Kepler's 3 laws.

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Nice example of a dynamical model 'without equations'.

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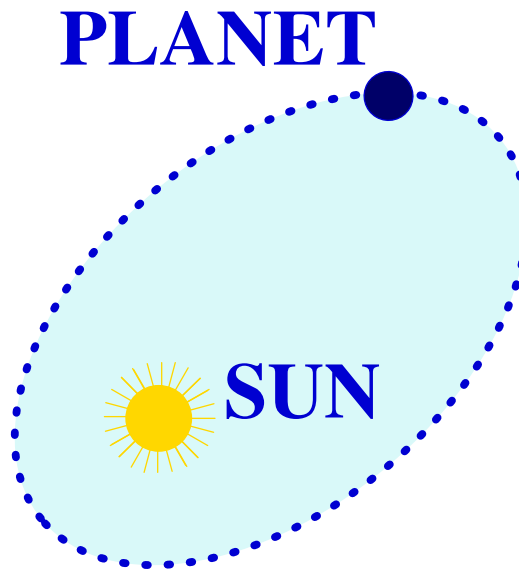
Nice example of a dynamical model 'without equations'.

*Is it a differential system?*

This question turned out to be of revolutionary importance...

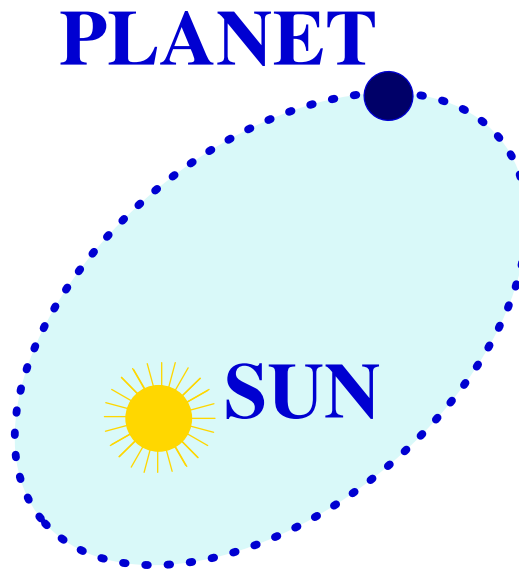
# The state of the planet

What determines the orbit uniquely?



# The state of the planet

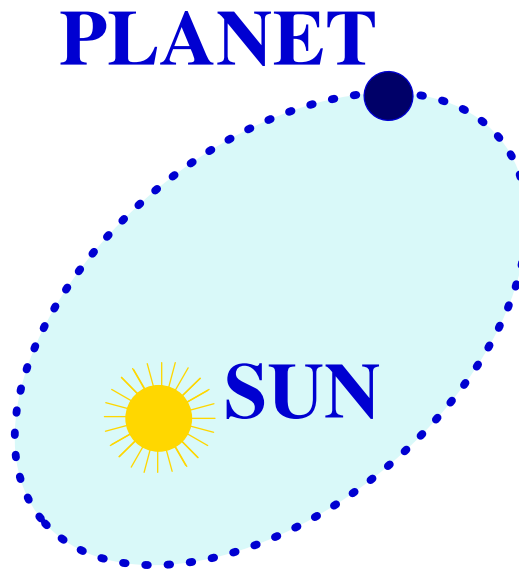
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The position?

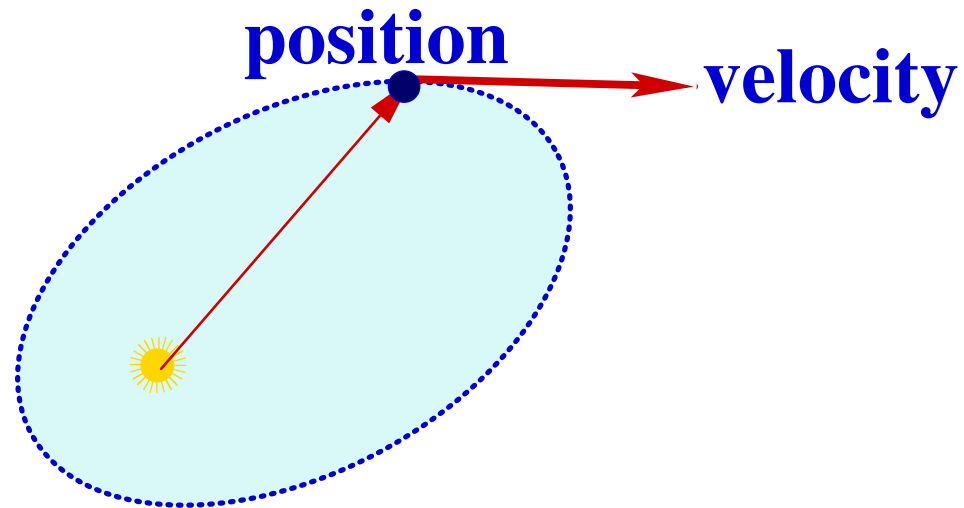
# The state of the planet

What determines the orbit uniquely?



The position and the direction of motion?

# The state of the planet



**The state = position and velocity**

# The equation of the planet

**Consequence:**

**acceleration = function of position and velocity**

$$\frac{d^2}{dt^2}w(t) = A\left(w(t), \frac{d}{dt}w(t)\right)$$



# The equation of the planet

Consequence:

**acceleration = function of position and velocity**

$$\frac{d^2}{dt^2}w(t) = A(w(t), \frac{d}{dt}w(t))$$

~> via **calculus** and **calculation**

$$\frac{d^2}{dt^2}w(t) + \frac{1_{w(t)}}{|w(t)|^2} = 0$$

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*Is it really equivalent to K.1, K.2, K.3 ?*



# Kepler's laws

**K.1, K.2, & K.3**

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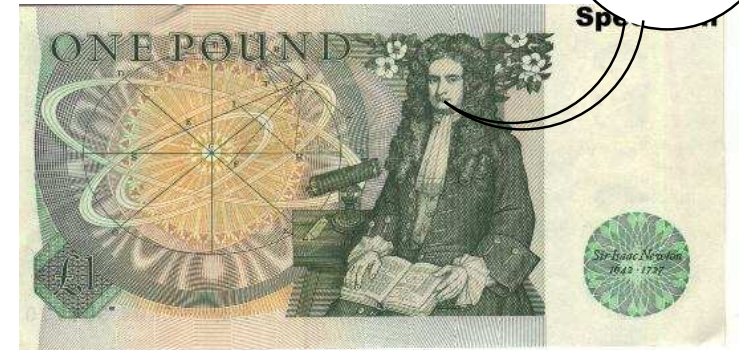
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Hypotheses  
non  
fingo





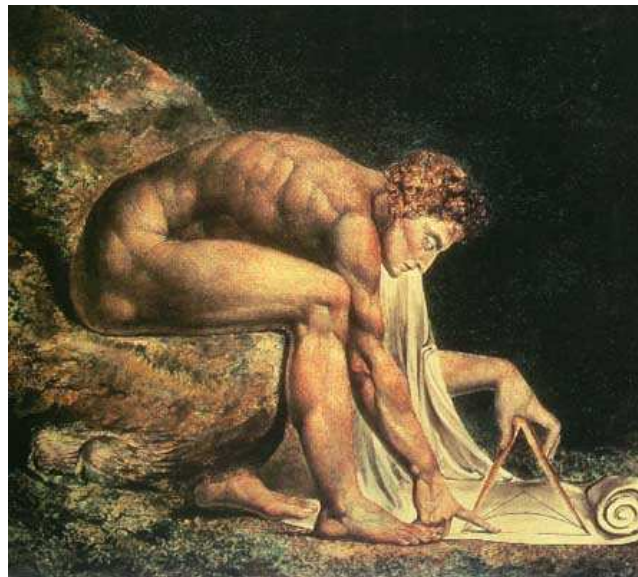
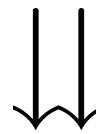
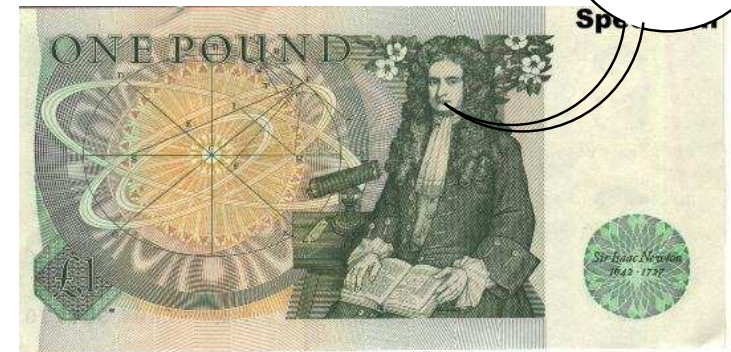
# Kepler's laws

**K.1, K.2, & K.3**



$$\frac{d^2}{dt^2}w(t) + \frac{1}{|w(t)|^2}w(t) = 0$$

Hypotheses  
non  
fingo



$$F = m \frac{d^2}{dt^2}w$$

## More examples

### Flows

$$\frac{d}{dt}x(t) = f(x(t)),$$

$\mathcal{B}$  = all state trajectories.



## More examples

### Observed flows

$$\frac{d}{dt}x(t) = f(x(t)); \quad y(t) = h(x(t)),$$

$\mathfrak{B}$  = all possible output trajectories.

### Note:

1. It may be impossible to express  $\mathfrak{B}$  as the solutions of a differential equation involving only  $y$ .
2. The auxiliary (latent variable) nature of  $x$ .

## More examples

### Input / output systems

$$f_1\left(y(t), \frac{d}{dt}y(t), \frac{d^2}{dt^2}y(t), \dots, t\right) \\ = f_2\left(u(t), \frac{d}{dt}u(t), \frac{d^2}{dt^2}u(t), \dots, t\right)$$

$T = \mathbb{R}$  (time),

$W = U \times Y$  (input  $\times$  output signal spaces),

$\mathcal{B} =$  all input / output pairs.



## More examples

### Input / output systems

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Of course, more is required to justify calling

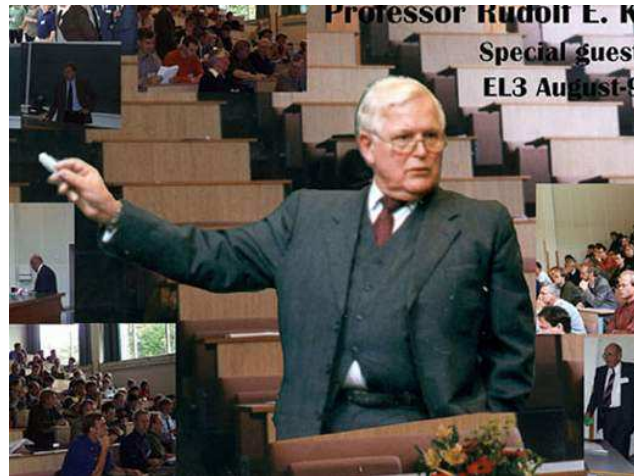
$u$  'input' and  $y$  'output'.

Not a good starting point!! **Why not?**

## More examples

### Input / state / output systems

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



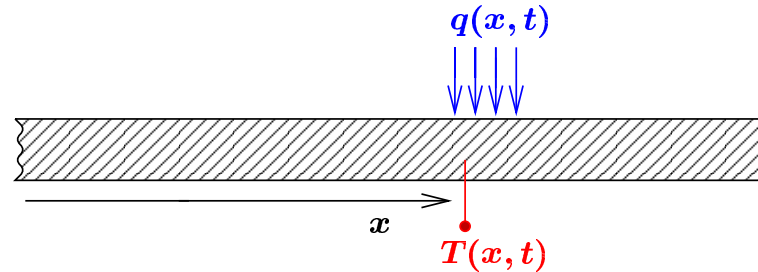
What do we want to call the behavior?

the  $(\mathbf{u}, \mathbf{y}, \mathbf{x})$ 's, or the  $(\mathbf{u}, \mathbf{y})$ 's?

Is the  $(\mathbf{u}, \mathbf{y})$  behavior described by a differential eq'n?

## More examples

### Heat diffusion



**Diffusion** describes the evolution of the **temperature**  $T(x, t)$  ( $x \in \mathbb{R}$  position,  $t \in \mathbb{R}$  time) along a uniform bar (infinitely long), and the **heat**  $q(x, T)$  supplied to the bar.  $\leadsto$  the PDE

$$\frac{\partial}{\partial t} T = \frac{\partial^2}{\partial x^2} T + q$$

$\mathbb{T} = \mathbb{R}$  (time),

$\mathbb{W} = \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^2)$  all (temperature, heat) distributions,

$\mathcal{B} =$  all  $T(\cdot, t), q(\cdot, t)$ -pairs that satisfy the PDE.

**Note:** We view  $t$  as a distinguished variable.

## More examples

### Codes

$\mathbb{A}$  = the code alphabet, say,  $\mathbb{A} = \mathbb{F}^w$ ,  $\mathbb{F}$  a finite field,

$\mathbb{I}$  = an index set, say,

$\mathbb{I} = (1, \dots, n)$  in block codes,

$\mathbb{I} = \mathbb{N}$  or  $\mathbb{Z}$  in convolutional codes,

$\mathcal{C} \subseteq \mathbb{A}^{\mathbb{I}}$  = **the code**; yields the system  $\Sigma = (\mathbb{I}, \mathbb{A}, \mathcal{C})$ .

Redundancy structure, error correction possibilities, etc., are visible in the code behavior  $\mathcal{C}$ .

**It is the central object of study.**

The encoder and decoder can be put (temporarily) into the background.

## More examples

### Example:

The following error detecting code:

$$\mathbb{I} = \mathbb{Z}, \mathbb{A} = \mathbb{F} = \{0, 1\},$$

$\mathcal{B}$  = all compact support sequences  $w : \mathbb{Z} \rightarrow \mathbb{F}$  such that

$$w(t) = p_0 \ell(t) + p_1 \ell(t - 1) + \cdots + p_n \ell(t - n)$$

for some  $\ell : \mathbb{Z} \rightarrow \mathbb{F}$ , with  $p_0, p_1, \dots, p_n \in \mathbb{F}$  design parameters.

## More examples

### *Formal languages*

$\mathbb{A}$  = a (finite) alphabet,

$\mathcal{L} \subseteq \mathbb{A}^*$  = **the language** = all 'legal' 'words'  $a_1 a_2 \cdots a_k \cdots$

$\mathbb{A}^*$  = all finite strings with symbols from  $\mathbb{A}$ .

yields the system  $\Sigma = (\mathbb{N}, \mathbb{A}, \mathcal{L})$ .

Examples: All words appearing in the *van Dale*  
All  $\text{L}^{\text{A}}\text{T}_{\text{E}}\text{X}$  documents.



## More examples

### Thermodynamics

***Thermodynamics is the only theory of a general nature of which I am convinced that it will never be overthrown.***

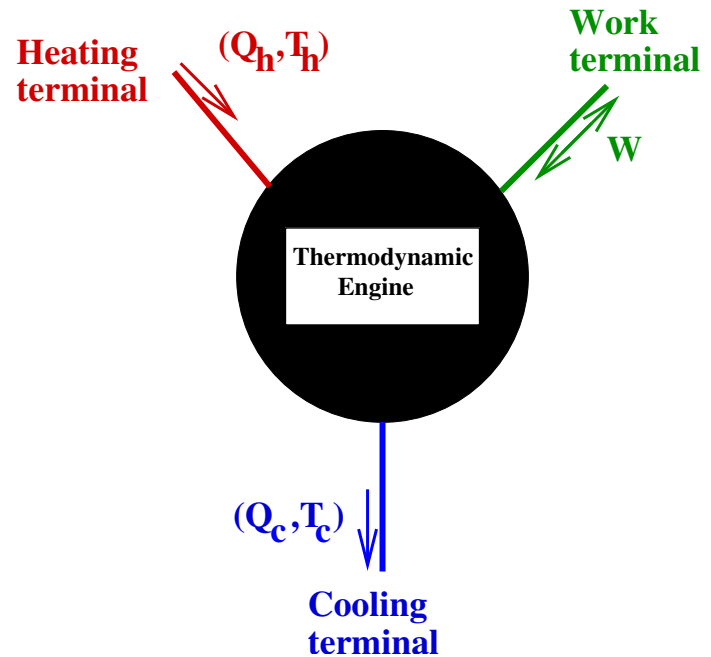
**Albert Einstein**

***The law that entropy always increases – the second law of thermodynamics – holds, I think, the supreme position among the laws of nature.***

**Arthur Eddington**

## More examples

### Thermodynamics



**time-axis:**  $\mathbb{R}$

**Q:** Variables of interest? **A:**  $Q_h, T_h, Q_c, T_c, W$

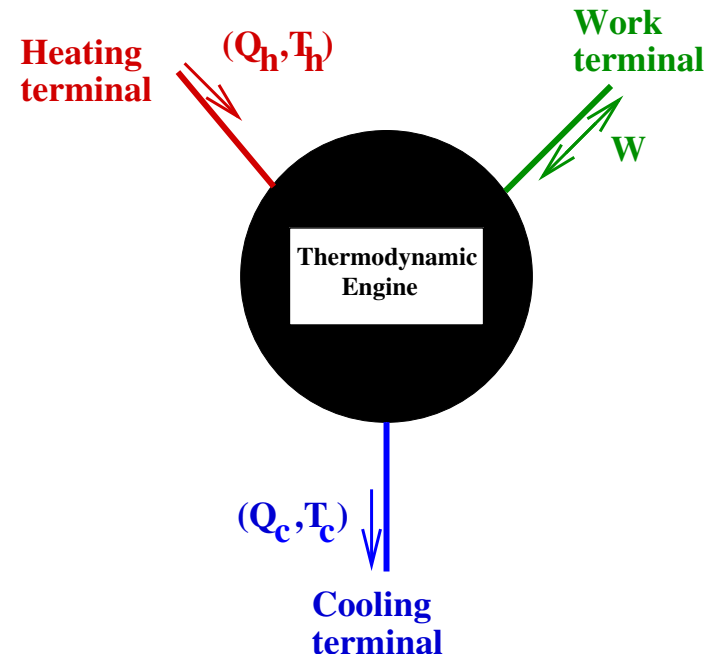
$\rightsquigarrow$  **signal space:**  $W = \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}$

**Behavior  $\mathcal{B}$ :** a suitable family of trajectories.

But, there are some universal laws that restrict the  $\mathcal{B}$ 's that are 'thermodynamic'.

## More examples

### Thermodynamics



First and second law:

$$\oint (Q_h - Q_c - W) dt = 0; \quad \oint \left( \frac{Q_h}{T_h} - \frac{Q_c}{T_c} \right) dt \leq 0.$$

These laws deal with **'open'** systems.

But not with input/output systems!

## Conclusion

**The framework**

$$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$$

**with  $\mathfrak{B}$  = the 'behavior'**

**is a very flexible, universal, pedagogical approach to the theory of models in general and dynamical systems in particular!**

# Linear Differential Systems

# Polynomials

$\mathbb{R}[\xi]$  := the polynomials with real coefficients in the 'indeterminate'  $\xi$ .

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Addition, multiplication defined in the obvious way.

This makes  $\mathbb{R}[\xi]$  into a ‘ring’.

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Addition, multiplication by polynomials (‘scalar multiplication’) defined in the obvious way.

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A **submodule** of  $\mathbb{R}^n[\xi]$  is a subset of  $\mathbb{R}^n[\xi]$  that is closed under addition and scalar multiplication.

# Polynomials

$\mathbb{R}[\xi]$  := the polynomials with real coefficients in the ‘indeterminate’  $\xi$ .

A **polynomial vector** is a vector of polynomials,  $\rightsquigarrow \mathbb{R}^n[\xi]$

Example of a submodule: Let  $g_1, g_2, \dots, g_k \in \mathbb{R}^n[\xi]$  and consider all elements of the form

$$p_1 g_1 + p_2 g_2 + \dots + p_k g_k$$

with  $p_1, p_2, \dots, p_k \in \mathbb{R}[\xi]$ . This is obviously a submodule of  $\mathbb{R}^n[\xi]$ , called the submodule ‘generated by’  $g_1, g_2, \dots, g_k$  (‘the generators’).

Notation:

$$\langle g_1, g_2, \dots, g_k \rangle \quad \text{or} \quad \langle G \rangle$$

if we organize the  $g$ ’s into the matrix  $G = \begin{bmatrix} g_1 & g_2 & \dots & g_k \end{bmatrix}$

# Polynomials

$\mathbb{R}[\xi]$  := the polynomials with real coefficients in the ‘indeterminate’  $\xi$ .

A **polynomial vector** is a vector of polynomials,  $\rightsquigarrow \mathbb{R}^n[\xi]$

Example of a submodule: Let  $g_1, g_2, \dots, g_k \in \mathbb{R}^n[\xi]$  and consider all elements of the form

$$p_1g_1 + p_2g_2 + \dots + p_kg_k$$

with  $p_1, p_2, \dots, p_k \in \mathbb{R}[\xi]$ . This is obviously a submodule of  $\mathbb{R}^n[\xi]$ , called the submodule ‘generated by’  $g_1, g_2, \dots, g_k$  (‘the generators’).

Every submodule of  $\mathbb{R}^n[\xi]$  is actually of this form: ‘finitely generated’; max. # of generators required is, in fact,  $n$ .

**Exercise: Prove this.**

# Polynomials

$\mathbb{R}[\xi]$  := the polynomials with real coefficients in the 'indeterminate'  $\xi$ .

A ***polynomial vector*** is a vector of polynomials,  $\rightsquigarrow \mathbb{R}^n[\xi]$

A ***polynomial matrix*** is a matrix of polynomials.

Notation:  $\mathbb{R}^n[\xi], \mathbb{R}^\bullet[\xi], \mathbb{R}^{n_1 \times n_2}[\xi], \mathbb{R}^{\bullet \times n}[\xi], \mathbb{R}^{n \times \bullet}[\xi], \mathbb{R}^{\bullet \times \bullet}[\xi]$ .

Add and multiply as if they were ordinary matrices.

# Polynomials

In  $\mathbb{R}^{n \times n}[\xi]$  we can add and multiply elements:

it is a 'non-commutative' ring.

So, elements could be invertible (in  $\mathbb{R}^{n \times n}[\xi]$ !).

These are an important type of square polynomial matrices:

**Definition:**  $P \in \mathbb{R}^{n \times n}[\xi]$  is said to be *unimodular* if  $\exists$   
 $Q \in \mathbb{R}^{n \times n}[\xi]$  such that  $QP = I_{n \times n}$ .

This  $Q$  is denoted as  $P^{-1}$ . Easy:  $PP^{-1} = P^{-1}P = I_{n \times n}$

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Proposition:  $P \in \mathbb{R}^{n \times n}[\xi]$  is unimodular if and only if  
 $\det(P) = \alpha$ , with  $0 \neq \alpha \in \mathbb{R}$ .

# Polynomials

The following result is very useful in proofs. It shows that, by pre- and postmultiplying by unimodular matrices, polynomial matrices can be brought in *Smith form*, a simple, diagonal like form (reminiscent of the Jordan form).



## Polynomials

Let  $P \in \mathbb{R}^{n_1 \times n_2}[\xi]$ . There exist unimodular polynomial matrices  $U \in \mathbb{R}^{n_1 \times n_1}[\xi]$  and  $V \in \mathbb{R}^{n_2 \times n_2}[\xi]$  such that

$$UPV = \begin{bmatrix} \text{diag}(p_1, p_2, \dots, p_r) & 0_{r \times (n_2 - r)} \\ 0_{(n_1 - r) \times r} & 0_{(n_1 - r) \times (n_2 - r)} \end{bmatrix}$$

where  $r = \text{rank}(P)$ ,  $p_1, p_2, \dots, p_r \in \mathbb{R}[\xi]$ , and  $p_{k+1}$  is a factor of  $p_k$  for  $k = 1, 2, \dots, r - 1$ .

$p_1, p_2, \dots, p_r$  are called the *invariant factors* of  $P$ .

# Polynomials

## Exercises:

- 1. Prove the Smith form.** Hint: Consider a non-zero element of  $P$  of smallest degree. Assume that in the same row or column as this element there is a second non-zero element. Now use an elementary row or column operation (meaning: replace this second row or column by the sum of this row or column and a polynomial multiple of the first row or column) to create either a new zero element, or a non-zero element of lower degree. Show that this process ends in a finite number of steps, and that it ends if, up to a permutation of rows and columns, the matrix is in Smith form.
- 2. Use the Smith form to prove that every submodule of  $\mathbb{R}^n[\xi]$  is finitely generated, and that the matrix formed by the generators as columns can be taken to have full row rank, and hence that the number of generators is at most  $n$ .**

# Linear time-invariant differential systems

We discuss the fundamentals of the theory of dynamical systems

$$\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathcal{B})$$

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$$((w_1, w_2 \in \mathfrak{B}) \wedge (\alpha, \beta \in \mathbb{R})) \Rightarrow (\alpha w_1 + \beta w_2 \in \mathfrak{B});$$

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2. **time-invariant** meaning

$$((w \in \mathfrak{B}) \wedge (t \in \mathbb{R})) \Rightarrow (\sigma^t w \in \mathfrak{B})$$

3. **differential,** meaning

$\mathfrak{B}$  consists of the sol'ns of a system of differential eq'ns.

# Linear time-invariant differential systems

Variables:  $w_1, w_2, \dots, w_w$ , up to  $n$ -times differentiated,  $g$  equations.  $\rightsquigarrow$

$$\begin{array}{r} \sum_{j=1}^w R_{1,j}^0 w_j + \sum_{j=1}^w R_{1,j}^1 \frac{d}{dt} w_j + \dots + \sum_{j=1}^w R_{1,j}^n \frac{d^n}{dt^n} w_j = 0 \\ \sum_{j=1}^w R_{2,j}^0 w_j + \sum_{j=1}^w R_{2,j}^1 \frac{d}{dt} w_j + \dots + \sum_{j=1}^w R_{2,j}^n \frac{d^n}{dt^n} w_j = 0 \\ \vdots \\ \sum_{j=1}^w R_{g,j}^0 w_j + \sum_{j=1}^w R_{g,j}^1 \frac{d}{dt} w_j + \dots + \sum_{j=1}^w R_{g,j}^n \frac{d^n}{dt^n} w_j = 0 \end{array}$$

Coefficients  $R_{1,j}^k$ : 3 indices!

$1 = 1, \dots, g$  : for the  $1$ -th differential equation,

$j = 1, \dots, w$  : for the variable  $w_j$  involved,

$k = 1, \dots, n$  : for the order  $\frac{d^k}{dt^k}$  of differentiation.

# Linear time-invariant differential systems

In vector/matrix notation:

$$w = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_w \end{bmatrix}, \quad R_k = \begin{bmatrix} R_{1,1}^k & R_{1,2}^k & \cdots & R_{1,w}^k \\ R_{2,1}^k & R_{2,2}^k & \cdots & R_{2,w}^k \\ \vdots & \vdots & \cdots & \vdots \\ R_{g,1}^k & R_{g,2}^k & \cdots & R_{g,w}^k \end{bmatrix} \cdot$$



# Linear time-invariant differential systems

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$$w = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_w \end{bmatrix}, \quad R_k = \begin{bmatrix} R_{1,1}^k & R_{1,2}^k & \cdots & R_{1,w}^k \\ R_{2,1}^k & R_{2,2}^k & \cdots & R_{2,w}^k \\ \vdots & \vdots & \cdots & \vdots \\ R_{g,1}^k & R_{g,2}^k & \cdots & R_{g,w}^k \end{bmatrix}.$$

Yields

$$R_0 w + R_1 \frac{d}{dt} w + \cdots + R_n \frac{d^n}{dt^n} w = 0,$$

with  $R_0, R_1, \cdots, R_n \in \mathbb{R}^{g \times w}$ .

# Linear time-invariant differential systems

**Yields**

$$R_0 w + R_1 \frac{d}{dt} w + \cdots + R_n \frac{d^n}{dt^n} w = 0,$$

with  $R_0, R_1, \dots, R_n \in \mathbb{R}^{g \times w}$ .

**Combined with the polynomial matrix**

$$R(\xi) = R_0 + R_1 \xi + \cdots + R_n \xi^n,$$

**we obtain the mercifully short notation**

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

**Other similar situations:**

**Difference equations on  $\mathbb{Z}_+$   $\rightsquigarrow$**

$$R(\sigma)w = 0;$$

**Difference equations on  $\mathbb{Z}$   $\rightsquigarrow$**

$$R(\sigma, \sigma^{-1})w = 0;$$

**PDE's on  $\mathbb{R}^n$   $\rightsquigarrow$**

$$R\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right)w = 0;$$

**Differential delay systems on  $\mathbb{R}$   $\rightsquigarrow$**

$$R\left(\frac{d}{dt}, \sigma, \sigma^{-1}\right)w = 0;$$

**etc.**

**Exercise: State and prove the appropriate version of the 3 basic theorems for discrete-time systems, with  $\mathbb{T} = \mathbb{Z}$  and  $\mathbb{Z}_+$ .**

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

*What do we mean by the behavior  
of this system of differential equations?*

When do we want to call  $w : \mathbb{R} \rightarrow \mathbb{R}^w$  a solution?

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When do we want to call  $w : \mathbb{R} \rightarrow \mathbb{R}^w$  a solution?

Possibilities:

**Strong** solutions?

**Weak** solutions?

$C^\infty(\mathbb{R}, \mathbb{R}^w)$  (infinitely differentiable) solutions?

**Distributional** solutions?

**Compact support** solutions, etc.

# The behavior of $R(\frac{d}{dt})w = 0$

## $\mathcal{C}^\infty$ -solutions:

$w : \mathbb{R} \rightarrow \mathbb{R}^w$  is a  $\mathcal{C}^\infty$ -solution of  $R(\frac{d}{dt})w = 0$  if

1.  $w$  is infinitely differentiable ( $=: w \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w)$ ), and
2.  $R(\frac{d}{dt})w = 0$ .

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## Weak solutions:

$w : \mathbb{R} \rightarrow \mathbb{R}^w$  is a weak solution of  $R\left(\frac{d}{dt}\right)w = 0$  if

1.  $\int_{t_0}^{t_1} \|w(t)\| dt < \infty$  for all  $t_0, t_1 \in \mathbb{R}$ , and
2.  $\int_{-\infty}^{+\infty} \left(R^\top\left(-\frac{d}{dt}\right)a\right)^\top(t)w(t) dt = 0$  for all  $a \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^{\text{rowdim}(R)})$  of compact support (i.e.,  $a$  is zero outside some finite interval).

# The behavior of $R(\frac{d}{dt})w = 0$

## $\mathcal{C}^\infty$ -solutions:

$w : \mathbb{R} \rightarrow \mathbb{R}^w$  is a  $\mathcal{C}^\infty$ -solution of  $R(\frac{d}{dt})w = 0$  if

1.  $w$  is infinitely differentiable ( $=: w \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w)$ ), and
2.  $R(\frac{d}{dt})w = 0$ .

## Weak solutions:

$w : \mathbb{R} \rightarrow \mathbb{R}^w$  is a weak solution of  $R(\frac{d}{dt})w = 0$  if

1.  $\int_{t_0}^{t_1} ||w(t)|| dt < \infty$  for all  $t_0, t_1 \in \mathbb{R}$ , and
2.  $\int_{-\infty}^{+\infty} (R^\top(-\frac{d}{dt})a)^\top(t)w(t) dt = 0$

‘Pragmatic’, easy way out:  $\mathcal{C}^\infty$  soln’s! Transmits main ideas, easier to handle, easy theory, sometimes (too) restrictive (step-response, state property etc.).



The behavior of  $R(\frac{d}{dt})w = 0$

Whence,  $R(\frac{d}{dt})w = 0$  defines the system  $\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathfrak{B})$  with

$$\mathfrak{B} = \{w \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w) \mid R(\frac{d}{dt})w = 0\}.$$

Proposition: This system is **linear** and **time-invariant**.

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### NOTATION

$\mathcal{L}^\bullet$  : all such systems

(with any - finite - number of variables)

$\mathcal{L}^w$  : with  $w$  variables

$$\mathfrak{B} = \ker(R(\frac{d}{dt}))$$

$\mathfrak{B} \in \mathcal{L}^w$  (no ambiguity regarding  $\mathbb{T}, \mathbb{W}$ )

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### NOMENCLATURE

Elements of  $\mathfrak{L}^\bullet$  : **linear differential systems**

$R(\frac{d}{dt})w = 0$  : a **kernel representation** of the  
corresponding  $\Sigma \in \mathfrak{L}^\bullet$  or  $\mathfrak{B} \in \mathfrak{L}^\bullet$

$R(\frac{d}{dt})w = 0$  'has' behavior  $\mathfrak{B}$

$\Sigma$  or  $\mathfrak{B}$ : the system **induced by**  $R \in \mathbb{R}^{\bullet \times \bullet}[\xi]$

# The behavior of $R(\frac{d}{dt})w = 0$

Some other properties of  $\mathfrak{B} \in \mathcal{L}^w$ :

$$(w \in \mathfrak{B}) \Rightarrow (\frac{d}{dt}w \in \mathfrak{B});$$

$$(w \in \mathfrak{B} \text{ and } p \in \mathbb{R}[\xi]) \Rightarrow (p(\frac{d}{dt})w \in \mathfrak{B});$$

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$$(w \in \mathfrak{B} \text{ and } p \in \mathbb{R}[\xi]) \Rightarrow (p(\frac{d}{dt})w \in \mathfrak{B});$$

Further niceties:

$$(w \in \mathfrak{B} \text{ and } f \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R})) \Rightarrow (f * w \in \mathfrak{B}),$$

\* denotes convolution

$\mathcal{C}^\infty$ -solutions of  $R(\frac{d}{dt})w = 0$  are **dense** in the set of weak (or distributional) solutions.

# 3 theorems about $\mathcal{L}^\bullet$

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### Theorem 1:

There is a one-to-one relation between  $\mathcal{L}^w$  and the submodules of  $\mathbb{R}^w[\xi]$ .

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### Theorem 2 (Elimination theorem):

$\mathcal{L}^\bullet$  is closed under projection.



## 3 theorems about $\mathcal{L}^\bullet$

### Theorem 1:

There is a one-to-one relation between  $\mathcal{L}^w$  and the submodules of  $\mathbb{R}^w[\xi]$ .

### Theorem 2 (Elimination theorem):

$\mathcal{L}^\bullet$  is closed under projection.

### Theorem 3 (Input / output representation):

For each  $\mathcal{B} \in \mathcal{L}^\bullet$  there is a componentwise partition of the variables into inputs and outputs.

This partition is not unique, but the numbers of input and output variables is invariant.

## Submodule theorem

$R \in \mathbb{R}^{\bullet \times \bullet}[\xi]$  defines  $\mathfrak{B} = \ker(R(\frac{d}{dt}))$ , but not vice-versa!

Obviously,  $R(\frac{d}{dt})w = 0$  and  $U(\frac{d}{dt})R(\frac{d}{dt})w = 0$  define the same behavior whenever  $U$  is unimodular.

??  $\exists$  'intrinsic' characterization of  $\mathfrak{B} \in \mathcal{L}^w$  ??

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??  $\exists$  'intrinsic' characterization of  $\mathfrak{B} \in \mathcal{L}^w$  ??

Is there a mathematical 'object' that characterizes a  $\mathfrak{B} \in \mathcal{L}^w$ ?

Define the **annihilators** of  $\mathfrak{B} \in \mathcal{L}^w$  by

$$\mathfrak{N}_{\mathfrak{B}} := \{n \in \mathbb{R}^w[\xi] \mid n^\top (\frac{d}{dt})\mathfrak{B} = 0\}.$$

Proposition:  $\mathfrak{N}_{\mathfrak{B}}$  is a  $\mathbb{R}[\xi]$  sub-module of  $\mathbb{R}^w[\xi]$ .

## Submodule theorem

$$\mathfrak{N}_{\mathfrak{B}} := \{n \in \mathbb{R}^w[\xi] \mid n^\top \left(\frac{d}{dt}\right)\mathfrak{B} = 0\}.$$

Let  $\langle R^\top \rangle$  denote the submodule of  $\mathbb{R}^w[\xi]$  spanned by the transposes of the rows of  $R$ . Obviously  $\langle R^\top \rangle \subseteq \mathfrak{N}_{\ker(R(\frac{d}{dt}))}$ .  
And, indeed:

$$\mathfrak{N}_{\ker(R(\frac{d}{dt}))} = \langle R^\top \rangle$$

**Note:** Depends on  $\mathcal{C}^\infty$ . False for compact support soln's.

## Submodule theorem

$$\mathfrak{N}_{\mathfrak{B}} := \{n \in \mathbb{R}^w[\xi] \mid n^\top \left(\frac{d}{dt}\right)\mathfrak{B} = 0\}.$$

$$\mathfrak{N}_{\ker(R(\frac{d}{dt}))} = \langle R^\top \rangle$$

Denote by  $\mathfrak{M}$  the set of submodules of  $\mathbb{R}^n[\xi]$ . Associate elements of  $\mathcal{L}^w$  with those of  $\mathfrak{M}$  as follows:

$$\mathfrak{B} \in \mathcal{L}^w \mapsto \mathfrak{N}_{\mathfrak{B}} \in \mathfrak{M}$$

$$M \in \mathfrak{M}, \mapsto \{w \mid n^\top (\text{der})w = 0 \forall n \in M\}$$

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Theorem 1:

$$\mathcal{L}^w \xleftrightarrow{1:1} \mathfrak{M}$$

## Submodule theorem

Consequence (**Structure of kernel representations**):

1. Let  $R_1\left(\frac{d}{dt}\right)w = 0$  have behavior  $\mathfrak{B}_1$ , and  $R_2\left(\frac{d}{dt}\right)w = 0$  have behavior  $\mathfrak{B}_2$ . Then  $\mathfrak{B}_1 \subseteq \mathfrak{B}_2$  if and only if  $\exists F \in \mathbb{R}^{\bullet \times \bullet}[\xi]$  such that  $R_2 = FR_1$ .
2.  $R\left(\frac{d}{dt}\right)w = 0$  is a **minimal kernel representation** (has a minimal number of rows over all kernel representations of a given behavior) if and only if  $R$  is of **full row rank**.
3. Let  $U$  be unimodular. Then  $R\left(\frac{d}{dt}\right)w = 0$  and  $U\left(\frac{d}{dt}\right)R\left(\frac{d}{dt}\right)w = 0$  have the same behavior.
4. Let  $R\left(\frac{d}{dt}\right)w = 0$  be minimal. **All minimal** kernel representations with the same behavior are obtained by **pre-multiplying**  $R$  by a **unimodular** polynomial matrix.

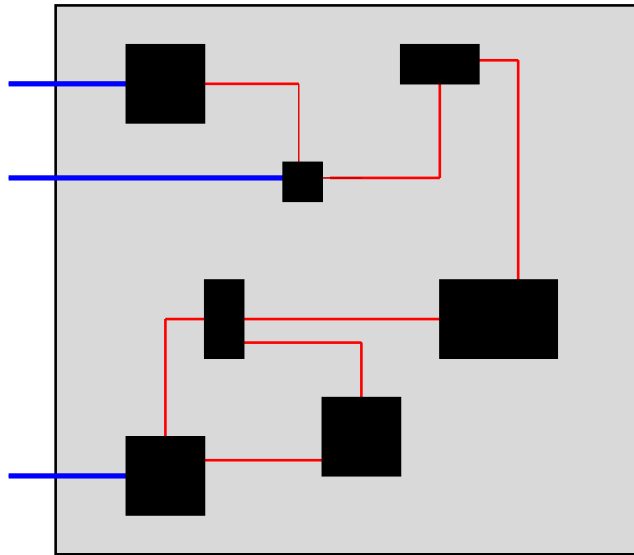
# Elimination theorem

**Motivation: In many problems, we want to eliminate variables.**



# Elimination theorem

Motivation: In many problems, we want to eliminate variables. For example, **first principle modeling**



~> model containing both variables the model aims at (**manifest** variables), and auxiliary variables introduced in the modeling process (**latent** variables).

*¿ Can these variables be eliminated from the equations ?*

## Elimination theorem

This leads to the following important question, first in polynomial matrix language. Consider

$$R_1\left(\frac{d}{dt}\right)w_1 = R_2\left(\frac{d}{dt}\right)w_2.$$

Obviously, the behavior of the  $(w_1, w_2)$ 's is described by a system of differential equation.

¿ But is the behavior of the  $w_1$ 's alone also ?

## Elimination theorem

In the language of behaviors:

Let  $\mathfrak{B} \in \mathcal{L}^{w_1 + w_2}$ . Define

$$\mathfrak{B}_1 = \{w_1 \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^{w_1}) \mid \exists w_2 \text{ such that } (w_1, w_2) \in \mathfrak{B}\}.$$

Does  $\mathfrak{B}$  belong to  $\mathcal{L}^{w_1}$  ?

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Does  $\mathfrak{B}$  belong to  $\mathcal{L}^{w_1}$  ?

Theorem 2: It does!

$\mathcal{L}^\bullet$  is closed under projection !!

Proof: follows from the ‘fundamental principle’.

Exercise 1: Prove the elimination th’ m, using the fund. pr.

Exercise 2: Prove the fund. pr. for const. coeff. lin. ODE’s.

# Elimination theorem

Example: ► The ubiquitous

$$\frac{d}{dt}\mathbf{x} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}; \quad \mathbf{y} = \mathbf{C}\mathbf{x} + \mathbf{D}\mathbf{u}, \quad \mathbf{w} = (\mathbf{u}, \mathbf{y}).$$

Which eq'ns describe the  $(\mathbf{u}, \mathbf{y})$  (input-output) behavior?

Elimination theorem  $\Rightarrow$  it is a system of differential eq'ns:

$$P\left(\frac{d}{dt}\right)\mathbf{y} = Q\left(\frac{d}{dt}\right)\mathbf{u}$$

with  $P$  square and  $\det(P) \neq 0$ . Why the latter?: soon!

# Elimination theorem

Example: ► The descriptor system

$$\frac{d}{dt}Ex + Fx + Gw = 0.$$

Which eqn's describe the  $w$  behavior?

Elimination theorem  $\Rightarrow$  it is a system of differential eq'ns:

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

! Compute  $(E, F, G) \mapsto R$ .

Row dimension of minimal kernel representation?

## Elimination theorem

Example: Consider a linear RLCTG circuit.

First principles modeling ( $\cong$  CE's, KVL, & KCL)

$\rightsquigarrow$  linear constant coefficient differential equations. All of them algebraic, only the L's and C's differential.

These include as variables both the **external port** and the **internal branch** voltages and currents.

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These include as variables both the **external port** and the **internal branch** voltages and currents.

**Can the port behavior be described by a system of linear constant coefficient differential equations?**

YES, because:

1. The CE's, KVL, & KCL are all linear constant coefficient differential equations.
2. The elimination theorem.

Row dimension of minimal kernel representation?

Interesting things can be said, using passivity! ...

## Elimination theorem

There exist effective algorithms for

$$(R_1, R_2) \mapsto R$$

incorporating, if desired, **minimality** of  $R(\frac{d}{dt})w = 0$ .

~> Computer algebra.

Start from  $R_1, R_2$ . We want to compute  $R$ .

Find a set of generators  $n_1, n_2, \dots, n_g$  for the **'left syzygy'** of the module  $\langle R_2 \rangle$ , i.e. for the module

$$\{n \in \mathbb{R}^{\text{coldim}(R_2)}[\xi] \mid n^\top R_2 = 0\}.$$

Define  $N = \begin{bmatrix} n_1 & n_2 & \cdots & n_g \end{bmatrix}$ . Then  $R = N^\top R_1$ .

## Elimination theorem

It follows from all this that  $\mathcal{L}^\bullet$  has very nice properties. In particular, it is **closed** under:

● **Intersection:**  $(\mathcal{B}_1, \mathcal{B}_2 \in \mathcal{L}^w) \Rightarrow (\mathcal{B}_1 \cap \mathcal{B}_2 \in \mathcal{L}^w).$

● **Addition:**  $(\mathcal{B}_1, \mathcal{B}_2 \in \mathcal{L}^w) \Rightarrow (\mathcal{B}_1 + \mathcal{B}_2 \in \mathcal{L}^w).$

● **Projection:**  $(\mathcal{B} \in \mathcal{L}^{w_1+w_2}) \Rightarrow (\Pi_{w_1} \mathcal{B} \in \mathcal{L}^{w_1}).$   
 $\Pi_{w_1} := \text{projection}$

● **Action of a linear differential operator:**

$$(\mathcal{B} \in \mathcal{L}^{w_1}, P \in \mathbb{R}^{w_2 \times w_1}[\xi]) \Rightarrow (P(\frac{d}{dt})\mathcal{B} \in \mathcal{L}^{w_2}).$$

● **Inverse image of a linear differential operator:**

$$(\mathcal{B} \in \mathcal{L}^{w_2}, P \in \mathbb{R}^{w_2 \times w_1}[\xi]) \Rightarrow (P(\frac{d}{dt}))^{-1}\mathcal{B} \in \mathcal{L}^{w_1}.$$

## I/O representation theorem

**When is a system variable an input? An output?**

# I/O representation theorem

When is a system variable an input? An output?

## Intuition

Our choice: the input is a **free variable** which, together with the **'initial conditions'**, determines the output.

# I/O representation theorem

When is a system variable an input? An output?

## Intuition

Our choice: the input is a **free variable** which, together with the **'initial conditions'**, determines the output.

These concepts (input, output) are strongly **domain dependent**. We follow the usual Systems & Control setting.

Central is, of course, that the input must in some way **cause** the output.

## I/O representation theorem

- In **real-time** signal processing and control, **non-anticipation** must be an important feature.
- But not in **non-real-time** signal processing problems, or when the independent variable is not time.
- In many problems (e.g. computing, signal processing) inputs may **have to be** structured, in order for machines or algorithms to be able to **accept** them.
- In control, it is customary to assume that inputs are free, and that outputs are bound (determined by the inputs and the initial conditions). We will follow this tradition.
- Often systems are non-anticipating **both forward and backward** in time.

## I/O representation theorem

$\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$ , with  $\mathbb{T} = \mathbb{R}$  or  $\mathbb{Z}$ , is said to be **autonomous** if

$$[w_1, w_2 \in \mathfrak{B}] \wedge [t \in \mathbb{T}] \wedge [w_1(t') = w_2(t') \quad \forall t' < t] \\ \Rightarrow [w_1 = w_2].$$

**Autonomous:= the past implies the future.**

**'Closed' systems.**



## I/O representation theorem

Let  $\Sigma = (\mathbb{R}, \mathbb{R}^{w_1+w_2}, \mathcal{B}) \in \mathcal{L}^{w_1+w_2}$ . Then  $w_1$  is said to be an **input/output system** with  $w_1$  the **input** and  $w_2$  the **output** if

1. the  $w_1$ -behavior  $\mathcal{B}_1 = \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^{w_1})$ , i.e.,  $w_1$  is **free**,
2. for all  $w_1 \in \mathcal{B}_1$ , the system  $\Sigma_2^{w_1} := (\mathbb{R}, \mathbb{R}^{w_2}, \mathcal{B}_2^{w_1})$  is **autonomous**,

where  $\mathcal{B}_2^{w_1}$  denotes the  $w_2$  behavior with fixed  $w_1$ , i.e.,

$$\mathcal{B}_2^{w_1} := \{w_2 \mid (w_1, w_2) \in \mathcal{B}\}.$$

**input**  $\cong$  **free**; **output**  $\cong$  **bound** (det. by input + initial cond's).

## I/O representation theorem

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**input**  $\cong$  **free**; **output**  $\cong$  **bound** (det. by input + initial cond's).

In keeping with tradition

$$w_1 \rightarrow u; \quad \mathbb{R}^{w_1} \rightarrow \mathbb{R}^m \text{ (m input variables),}$$

$$w_2 \rightarrow y; \quad \mathbb{R}^{w_2} \rightarrow \mathbb{R}^p \text{ (p output variables).}$$

## I/O representation theorem

Let  $\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathfrak{B}) \in \mathcal{L}^w$ , with  $\mathbb{R}^w = \mathbb{R}^m \times \mathbb{R}^p$ ,  $w = m + p$ .

If  $\Sigma = (\mathbb{R}, \mathbb{R}^m \times \mathbb{R}^p, \mathfrak{B})$  is an input/output system, then we call  $w = (u, y)$  an **input/output partition** of  $w$ .

# I/O representation theorem

## Notation for vectors and matrices of rational functions:

$$\mathbb{R}(\xi)^n, \mathbb{R}(\xi)^\bullet, \mathbb{R}(\xi)^{n_1 \times n_2}, \mathbb{R}(\xi)^{\bullet \times n}, \mathbb{R}(\xi)^{n \times \bullet}, \mathbb{R}(\xi)^{\bullet \times \bullet}.$$

A rational function  $\frac{a}{b} \in \mathbb{R}(\xi)$  is said to be

*proper* if  $\text{degree}(a) \leq \text{degree}(b)$ ,

*strictly proper* if  $\text{degree}(a) < \text{degree}(b)$ ,

and *bi-proper* if  $\text{degree}(a) = \text{degree}(b)$ .

$\rightsquigarrow$  vectors, matrices of (strictly) proper rational functions.

# I/O representation theorem

Proposition: Consider the kernel representation

$$P\left(\frac{d}{dt}\right)y = Q\left(\frac{d}{dt}\right)u, w = (u, y).$$

Then  **$u$  is free**  $\Leftrightarrow \text{rank}([P \ Q]) = \text{rank}(P),$

and  **$y$  is bound**  $\Leftrightarrow P$  is of full column rank,

i.e.  $\text{rank}(P) = \dim(y).$

It defines an **input/output partition** if and only if

$$\text{rank}([P \ Q]) = \text{rank}(P) = \dim(y).$$

If it is a minimal kernel representation, then I/O partition if and only if  **$P$  is square, and  $\det(P) \neq 0.$**

Call

$$G := P^{-1}Q$$

$\in \mathbb{R}(\xi)^{p \times m}$

its **transfer function.**

# I/O representation theorem

## Theorem 3:

Every system  $\Sigma \in \mathcal{L}^\bullet$  admits an input/output partition.

# I/O representation theorem

## Theorem 3:

Every system  $\Sigma \in \mathcal{L}^\bullet$  admits an input/output partition.

**even a componentwise I/O partition**

**:= some well-chosen components of  $w$  are the inputs,  
the others are the outputs**

**$\cong$  up to re-ordering of the variables,  $w = (u, y)$ ,  
i.e.,  $(u, y) = \Pi w$ , with  $\Pi$  a permutation matrix.**

**In fact,  $\exists$  a componentwise partition with  $G$  proper.**

**If one can choose the basis, and then the partition, even with  $G$  strictly proper.**

# I/O representation theorem

## Notes:

1. For a given  $\mathfrak{B} \in \mathfrak{L}^\bullet$ , which variables are input variables, and which are output variables, is not fixed.

**THIS IS A GOOD THING!**

## Examples:

An Ohmic resistor  $V = RI$   $R \neq 0$  may be viewed as a **current controlled** or as a **voltage controlled** device.

Often, a transfer function is bi-proper, and then there is no reason to prefer one input/output partition over another.

etc., etc.



## I/O representation theorem

2. The **number** of input and the **number** of output variables are fixed by  $\mathfrak{B} \in \mathcal{L}^\bullet$ .

## I/O representation theorem

**Notation:** Define the 3 maps  $w, m, p : \mathcal{L}^\bullet \rightarrow \mathbb{Z}_+$  by

$$w(\Sigma) = w(\mathcal{B}) := \text{the number of variables}$$

$$m(\Sigma) = m(\mathcal{B}) := \text{the number of input variables}$$

$$p(\Sigma) = p(\mathcal{B}) := \text{the number of output variables}$$

In terms of the kernel representation  $R(\frac{d}{dt})w = 0$ , we have

$$w(\Sigma) = \text{coldim}(R),$$

$$m(\Sigma) = \text{coldim}(R) - \text{rank}(R),$$

$$p(\Sigma) = \text{rank}(P)$$

In particular,  $m + p = w$ .

**End of the Lecture I**