



OPEN AND CONNECTED

Jan C. Willems
K.U. Leuven, Belgium

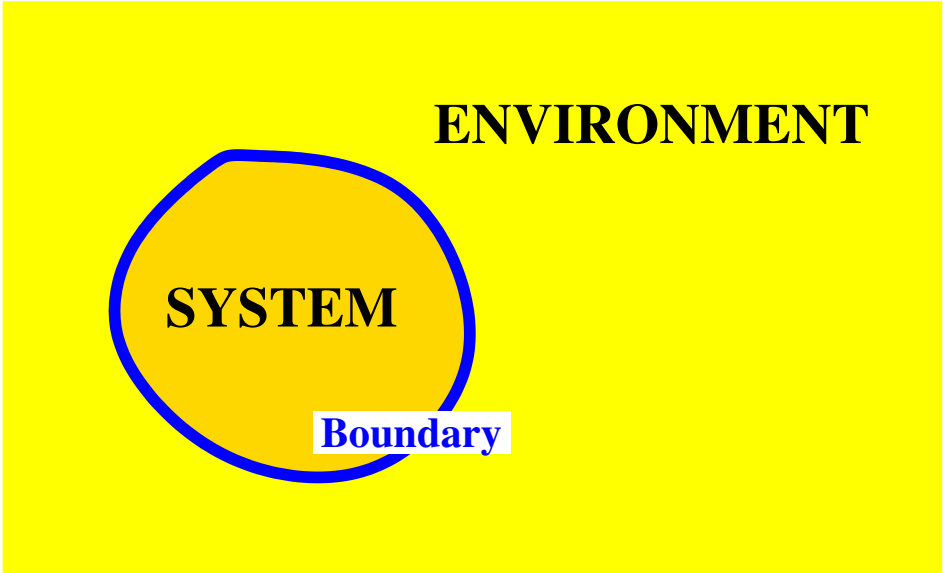
Open and Connected

The central tenets of our field:

Systems are **open** and consist of
interconnected subsystems.

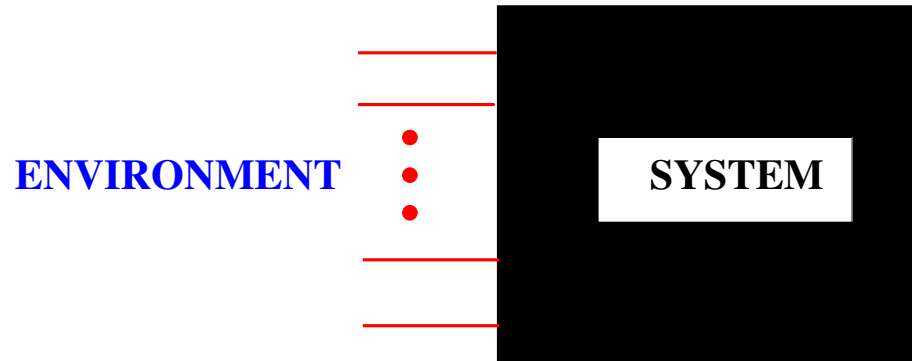
Synthesis of systems consists of
interconnecting subsystems

Open



Open

In this lecture, we think of this interaction boundary as 'terminals'



electrical components with '**wires**'

mechanical components with '**pins**'

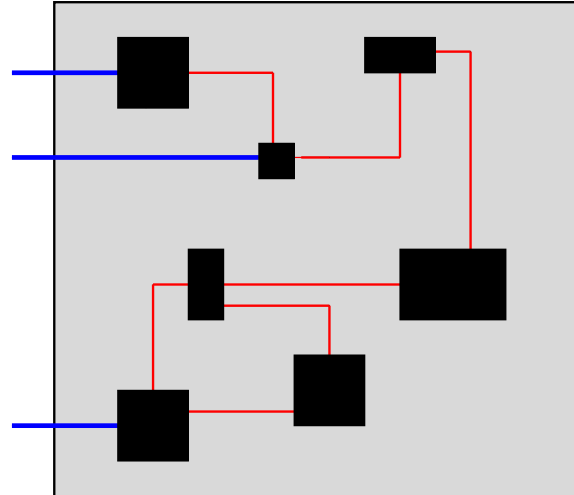
fluidic components with '**ducts**'

signal processors with **inputs and outputs**

motors with terminals & pins

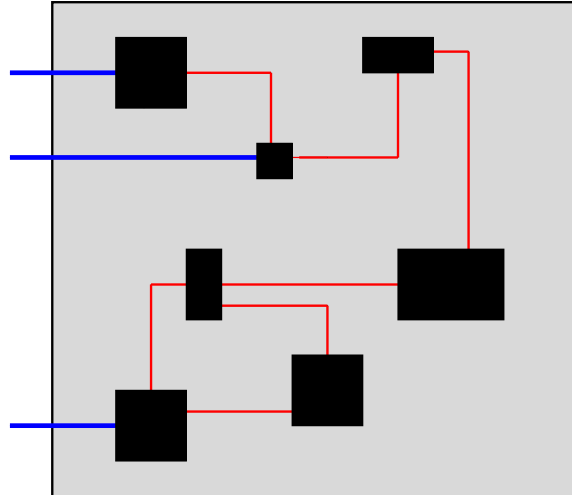
computer terminal, etc., etc., etc.

Connected



An interconnection architecture with subsystems

Connected



Think of:

electrical circuits

mechanical constructions

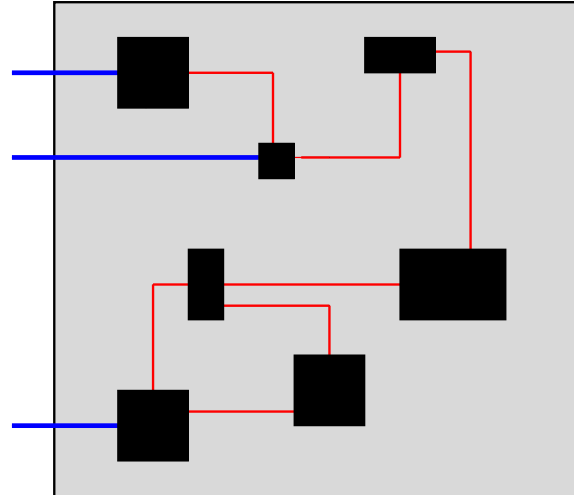
fluidic systems

networks of **signal processors**

computers

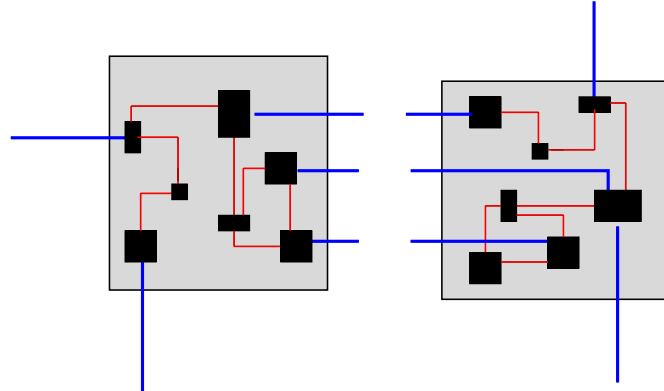
essentially all **engineering systems**

Connected

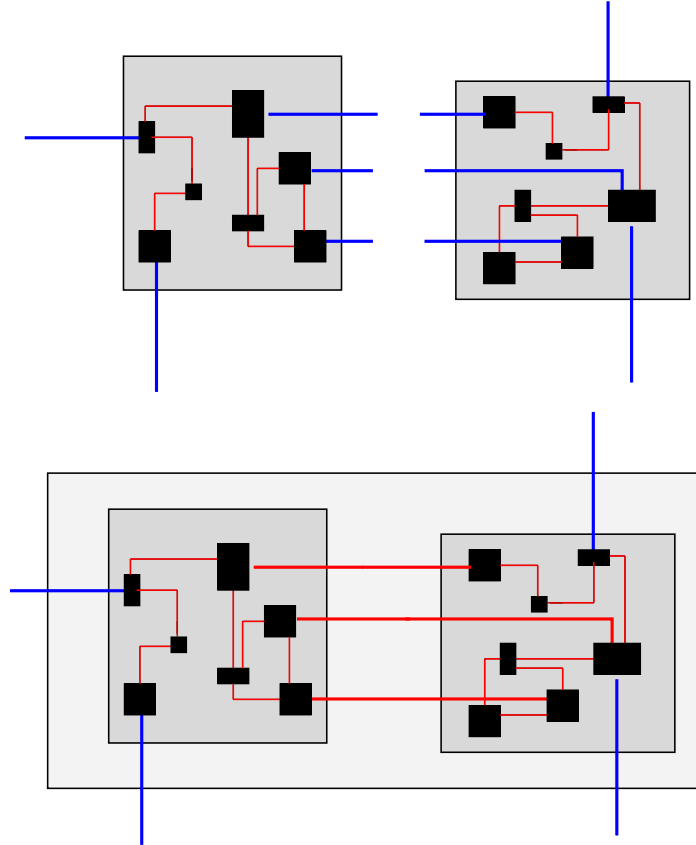


Observe the **hierarchical** nature

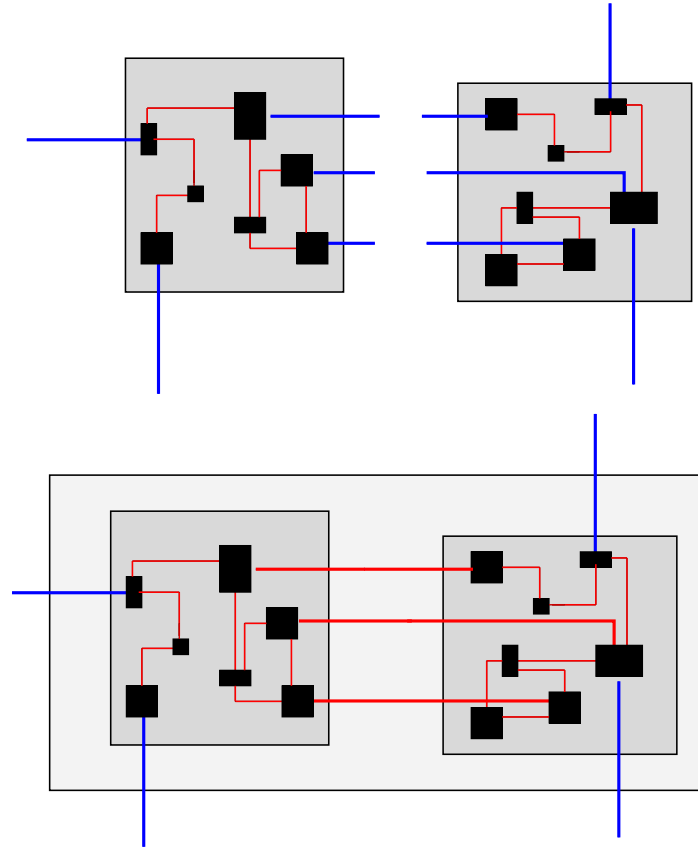
Interconnect



Interconnect



Interconnect



Reverse process: **'tearing'** & **'zooming'** & **'linking'**:

very useful in **modeling**.

Mathematization

What are the appropriate concepts / mathematization?

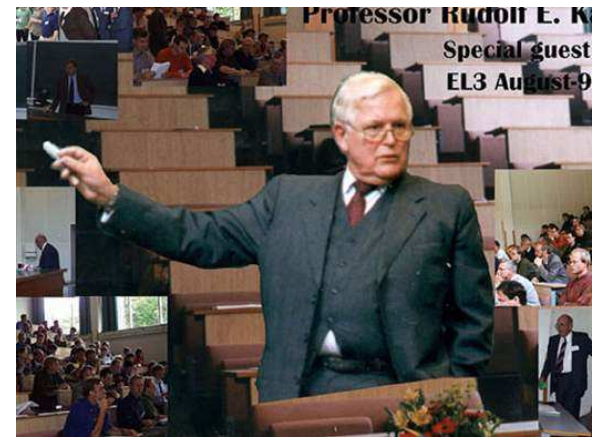
What is an **open** dynamical system?

How do we deal with **interconnections**?

How does **control** fit in?

Mathematization

1. **Get the physics right**
2. The rest is mathematics



**R.E. Kalman, Opening lecture
IFAC World Congress, Prague, July 4, 2005**

THEMES

1. Open and connected
2. A brief history of systems theory
3. **Control, interconnection, inputs and outputs**
4. Models and behaviors
5. Linear time-invariant differential systems
6. Controllability and stabilizability
7. **Representations of linear differential systems**
8. PDE's

The paradigm of closed systems

Axiomatization

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

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

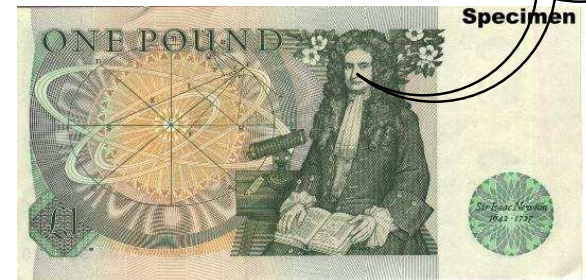
$$\rightsquigarrow \frac{d}{dt}x = f(x)$$

\rightsquigarrow 'dynamical systems', flows

\rightsquigarrow **closed systems** as paradigm of dynamics



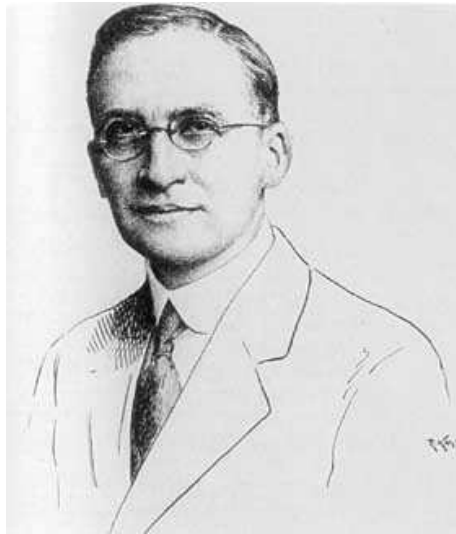
Hypotheses
non
fingo



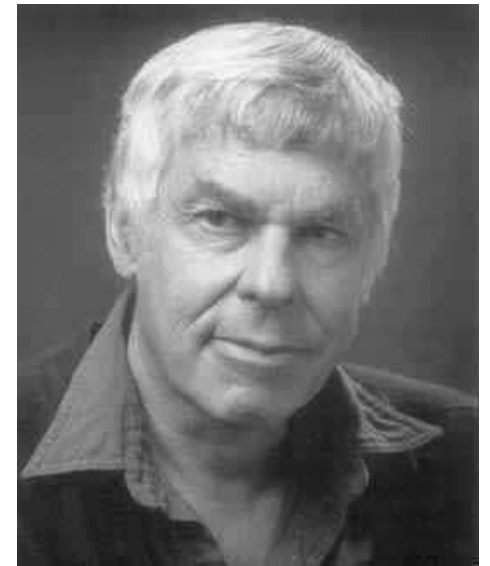
Axiomatization



Henri Poincaré (1854-1912)



George Birkhoff (1884-1944)



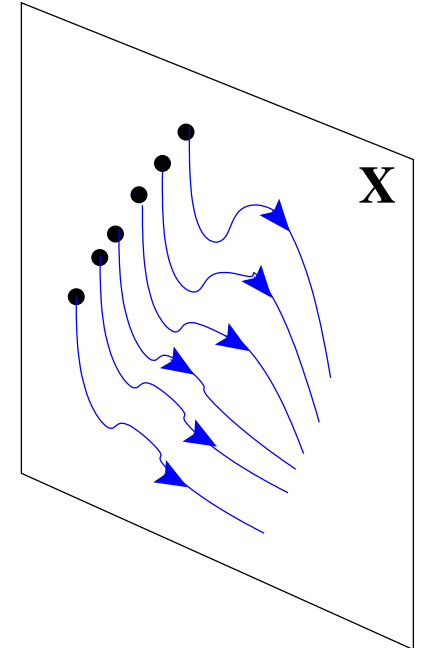
Stephen Smale (1930-)

Axiomatization

A *dynamical system* is defined by
a **state space** X and
a **state transition function**

$\phi : \dots$ such that \dots

$\phi(t, x)$ = state at time t starting from state x



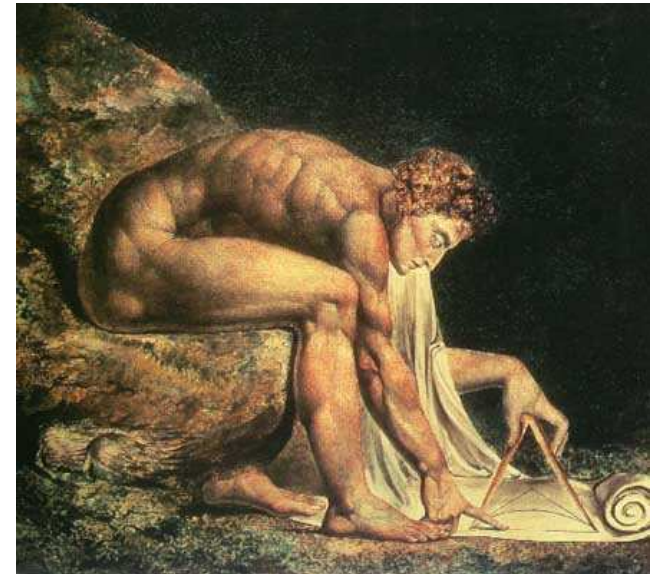
How could they forget about Newton's second law,
about Maxwell's eq'ns,
about thermodynamics,
about tearing & zooming & linking, ...?

Newton's laws

2-nd law $F'(t) = m \frac{d^2}{dt^2} w(t)$

gravity $F''(t) = m \frac{1_{w(t)}}{|w(t)|^2}$

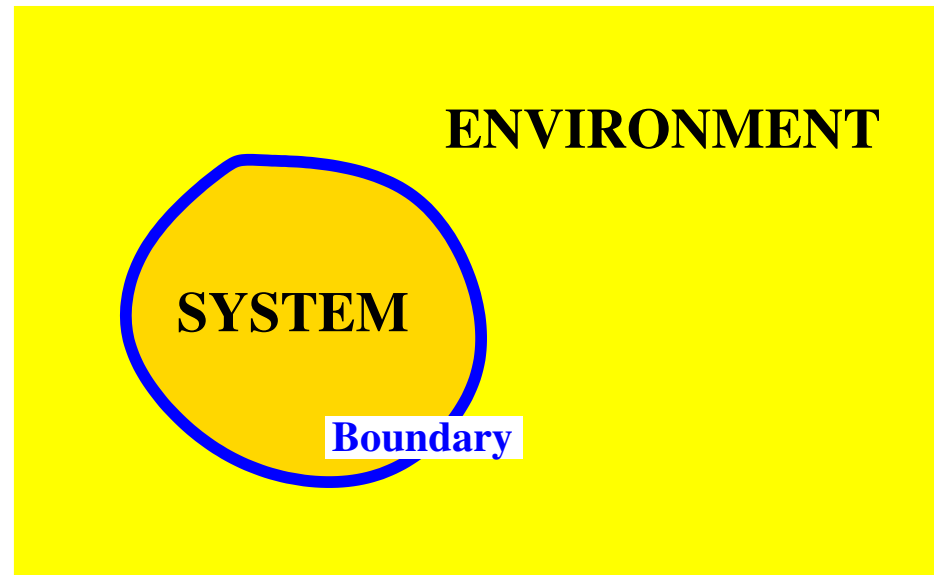
3-rd law $F'(t) + F''(t) = 0$



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

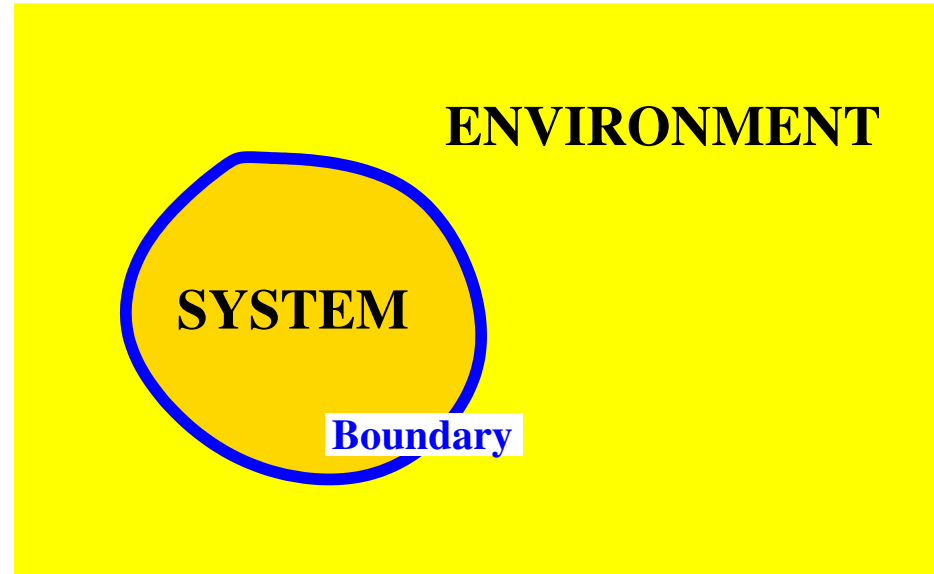
Closed systems

Reply: assume **'fixed boundary conditions'**



~> an absurd situation: to model a system,
we have to model also the environment!

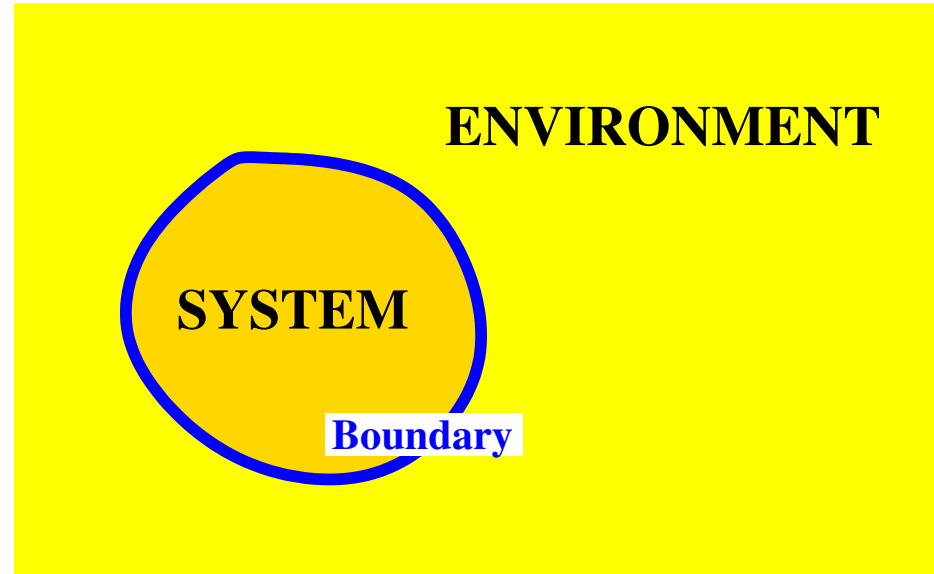
Closed systems



Chaos theory, cellular automata, sync, etc.,

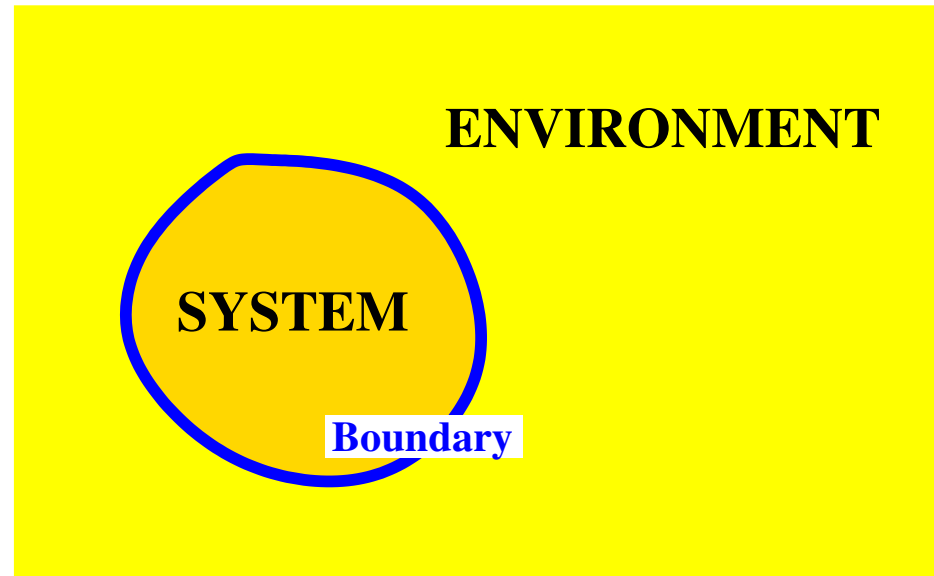
'function' in this framework ...

Closed systems



**Chaos: not a property of the physical laws,
but just as much of what the system is
interconnected to.**

Closed systems

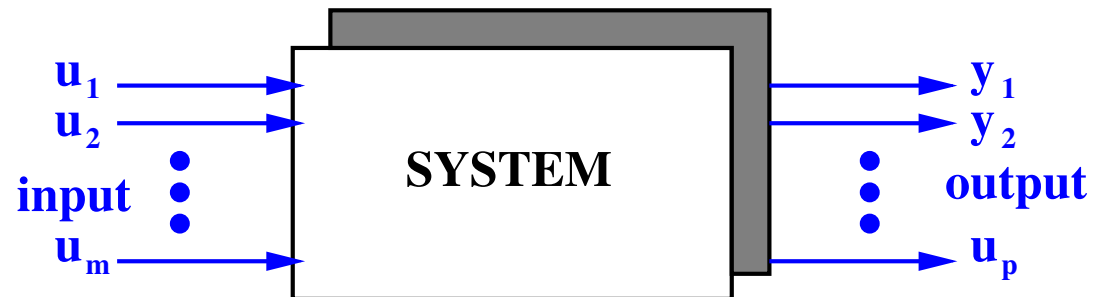
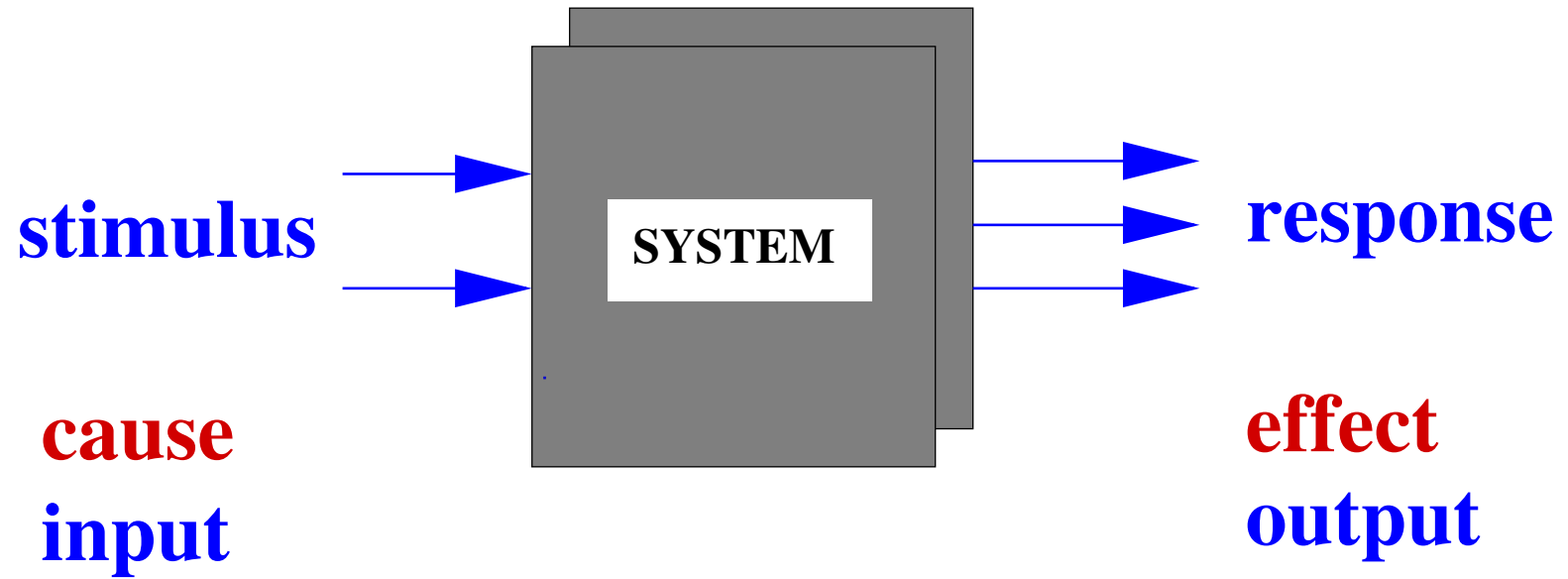


Turbulence may not be a property of Navier-Stokes, but just as much of the boundary conditions.

Meanwhile, in engineering, ...

The paradigm of input/output systems

Input/output systems



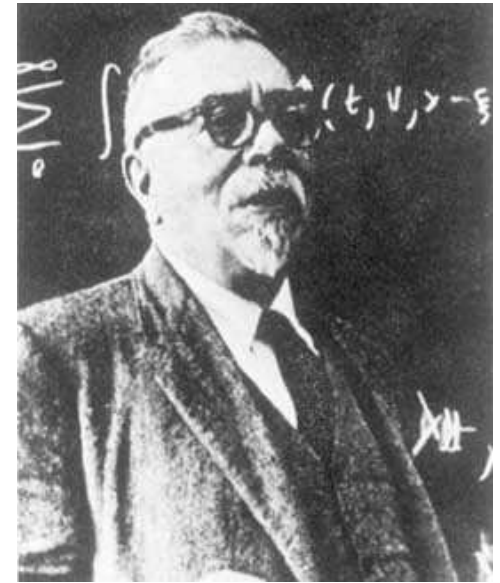
The originators



Lord Rayleigh (1842-1919)



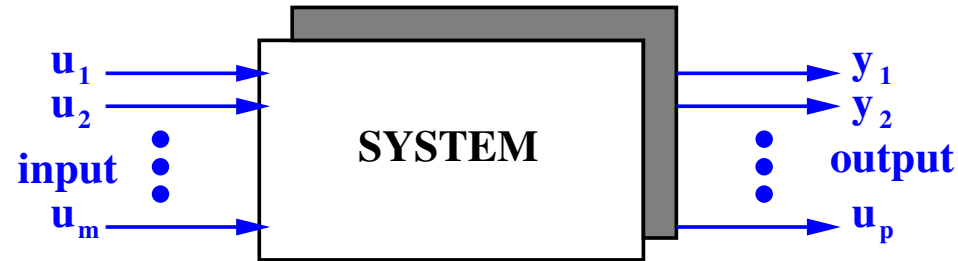
Oliver Heaviside (1850-1925)



Norbert Wiener (1894-1964)

and the many electrical circuit theorists ...

Mathematical description



$$y(t) = \int_{0 \text{ or } -\infty}^t H(t - t') u(t') dt'$$

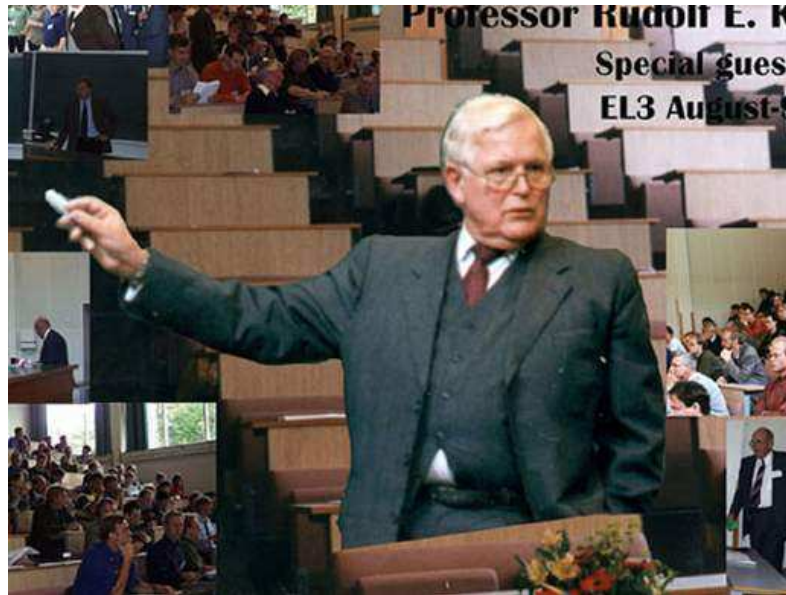
$$y(t) = H_0(t) + \int_{-\infty}^t H_1(t - t') u(t') dt' + \int_{-\infty}^t \int_{-\infty}^{t'} H_2(t - t', t' - t'') u(t') u(t'') dt' dt'' + \dots$$

These models fail to deal with **'initial conditions'**.

A physical system is **SELDOM** an i/o [map](#)

Input/state/output systems

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

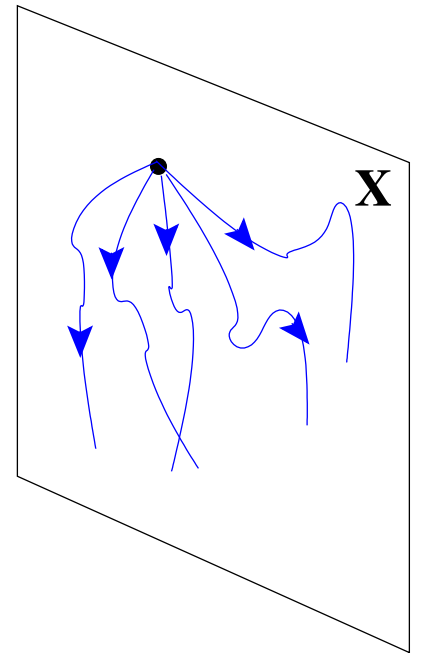


Rudolf Kalman (1930-)

'Axiomatization'

State transition function:

$\phi(t, x, u)$: state reached at time t from x using input u .

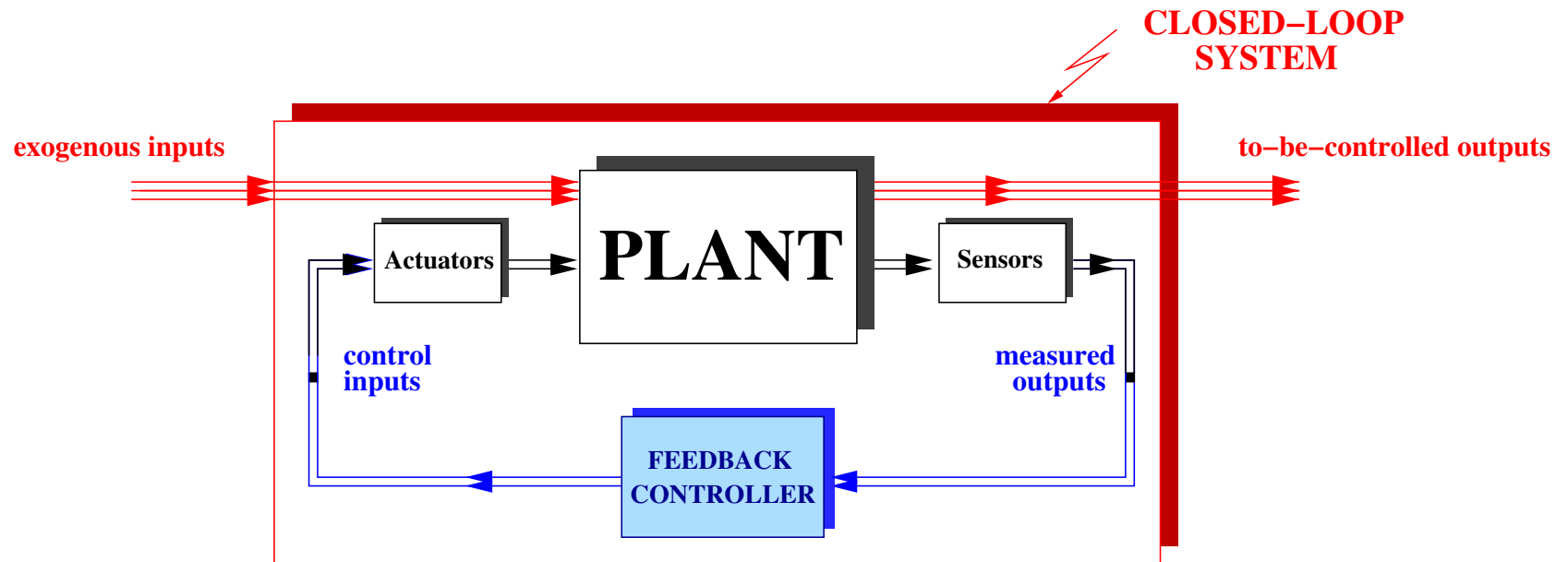


Read-out function:

$g(x, u)$: output value with state x and input value u .

The **input/state/output** view turned out to be
a very effective and fruitful paradigm

- for **control** (stabilization, robustness, ...)



The **input/state/output** view turned out to be
a very effective and fruitful paradigm

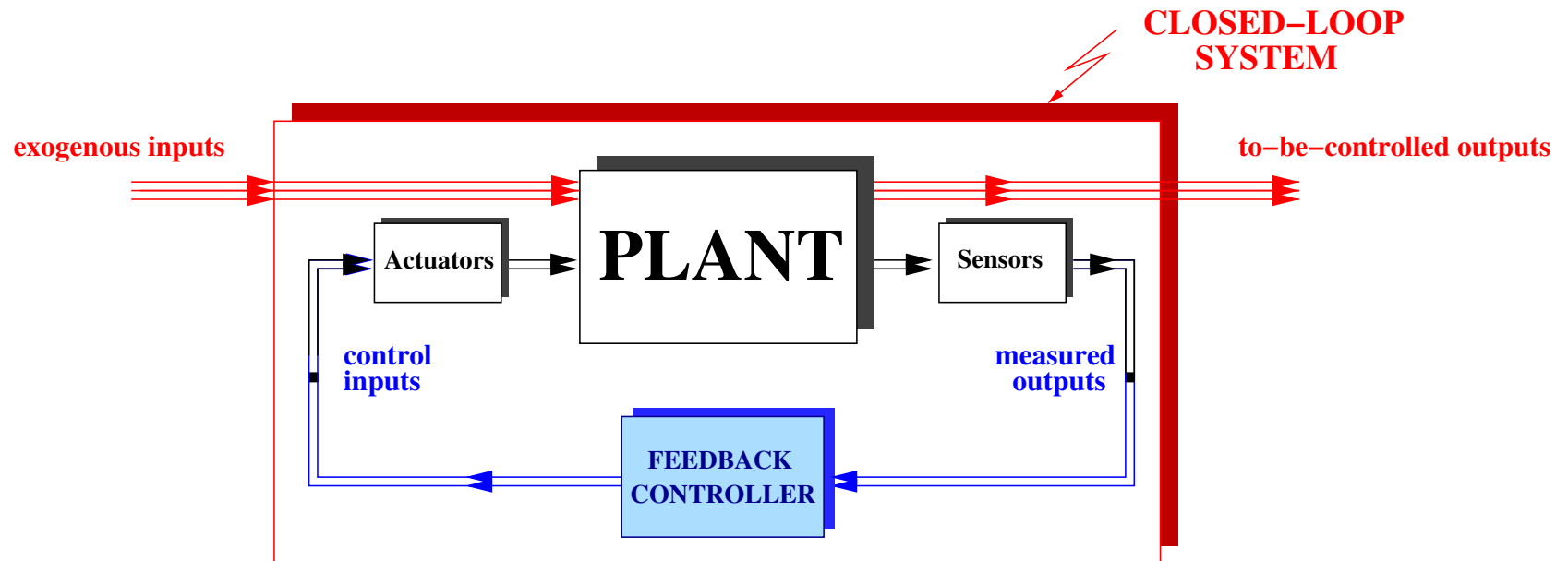
- for **control** (stabilization, robustness, ...)
- **prediction** of one signal from another, **filtering**
- understanding **system representations**
(transfer f'n, input/state/output, etc.)
- model simplification, **reduction**
- **system ID:** models from data
- etc., etc., etc.

Let's take a closer look at the i/o framework ...

in control

Difficulties with i/o

active control



versus **passive control**

Dampers, heat fins, pressure valves, ...

Controllers without sensors and actuators

Difficulties with i/o

active control versus passive control

Controlling turbulence

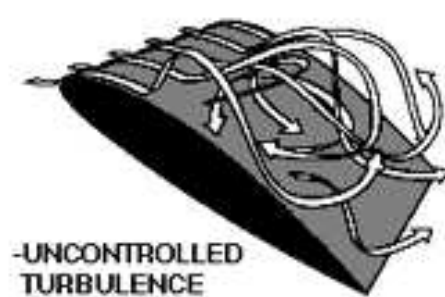
for airplanes, sharks, dolphins, golf balls, bicycling helmets, etc.



Difficulties with i/o

active control versus passive control

Controlling turbulence



Difficulties with i/o

active control versus passive control

Controlling turbulence

Nagano 1998



Difficulties with i/o

active control versus **passive** control

Controlling turbulence

Nagano 1998



Difficulties with i/o

active control versus passive control

Controlling turbulence

Nagano 1998

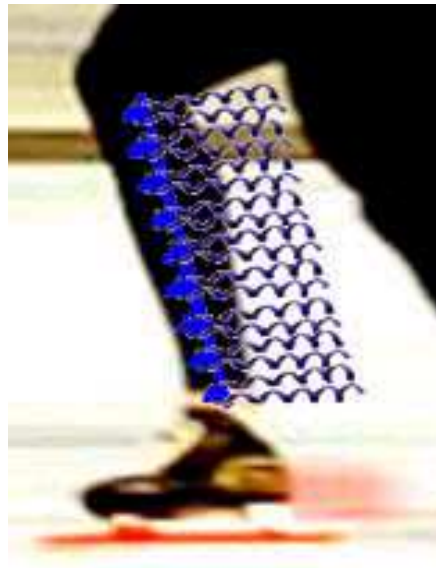


Difficulties with i/o

active control versus passive control

Controlling turbulence

Nagano 1998

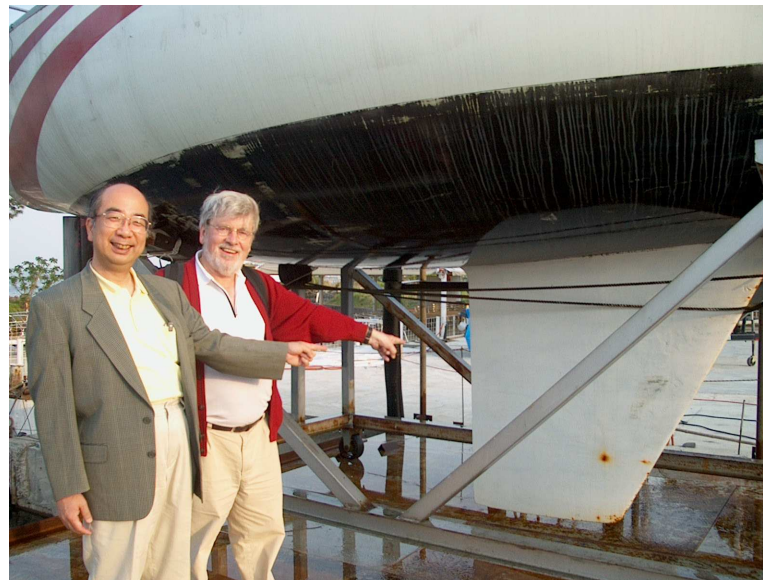


These are beautiful **controllers!** But, the only people not calling this "**control**", are the **control engineers** ...

Difficulties with i/o

active control versus **passive** control

Another example: the stabilizer of a ship



These are beautiful **controllers**! But, the only people not calling this **”stabilization”**, are the **control engineers** ...

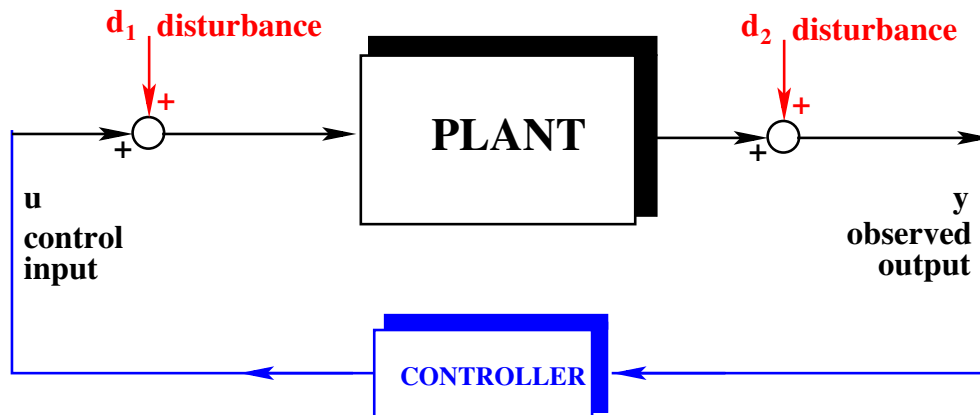
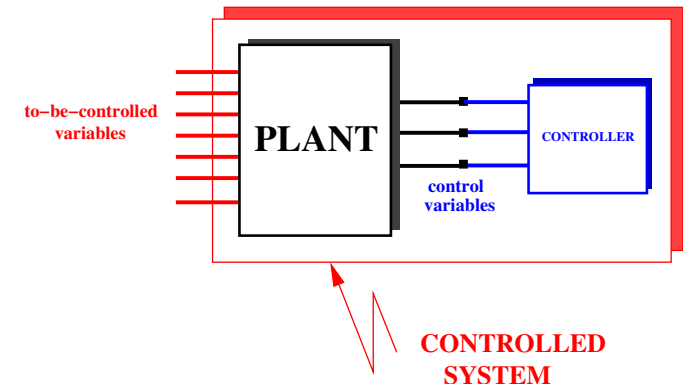
Btw, this interconnection is, but shouldn't be, called **‘singular’**

Difficulties with i/o

active control versus passive control

The appropriate figure is

With the 'classical' interconnection figure



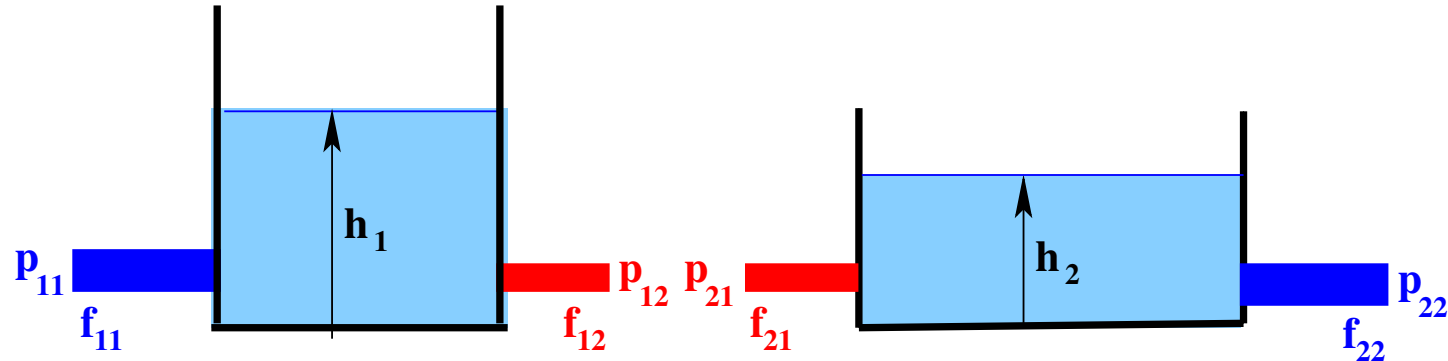
such controllers do not stabilize, because

dynamic order controlled system $<$ dynamic order plant $+$ dynamic order contro

Let's take a closer look at the i/o framework ...

for interconnection

i/o and interconnection



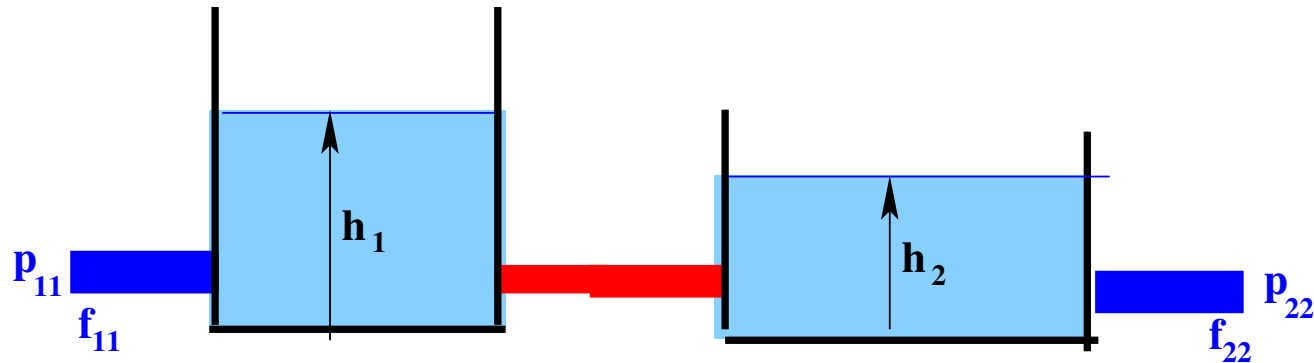
$$\frac{d}{dt}h_1 = F_1(h_1, p_{11}, p_{12}), f_{11} = H_{11}(h_1, p_{11}), f_{12} = H_{12}(h_1, p_{12})$$

$$\frac{d}{dt}h_2 = F_1(h_2, p_{21}, p_{22}), f_{21} = H_{21}(h_2, p_{21}), f_{22} = H_{22}(h_2, p_{22})$$

inputs: the pressures $p_{11}, p_{12}, p_{21}, p_{22}$

outputs: the flows $f_{11}, f_{12}, f_{21}, f_{22}$

i/o and interconnection



$$\frac{d}{dt}h_1 = F_1(h_1, p_{11}, p_{12}), f_{11} = H_{11}(h_1, p_{11}), f_{12} = H_{12}(h_1, p_{12})$$

$$\frac{d}{dt}h_2 = F_2(h_2, p_{21}, p_{22}), f_{21} = H_{21}(h_2, p_{21}), f_{22} = H_{22}(h_2, p_{22})$$

Interconnection:

$$p_{12} = p_{21}, f_{12} + f_{21} = 0$$

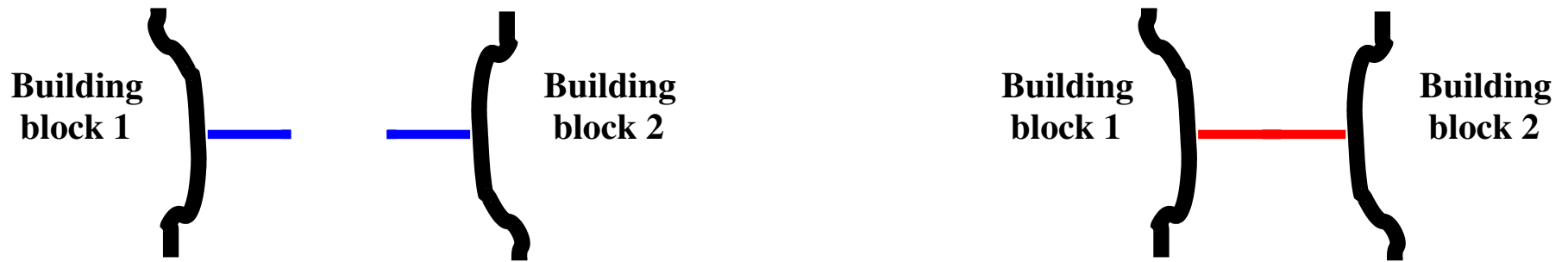
This identifies 2 inputs **AND (NOT WITH)** 2 outputs,

the sort of thing SIMULINK[©] forbids.

This situation is **the rule, not the exception** (in fluidics, mechanics,...)

Interconnection is not input-to-output assignment!

Sharing variables, not input-to-output assignment, is the basic mechanism by which systems interact.



Before interconnection:

the variables on the interconnected terminals are **independent**.

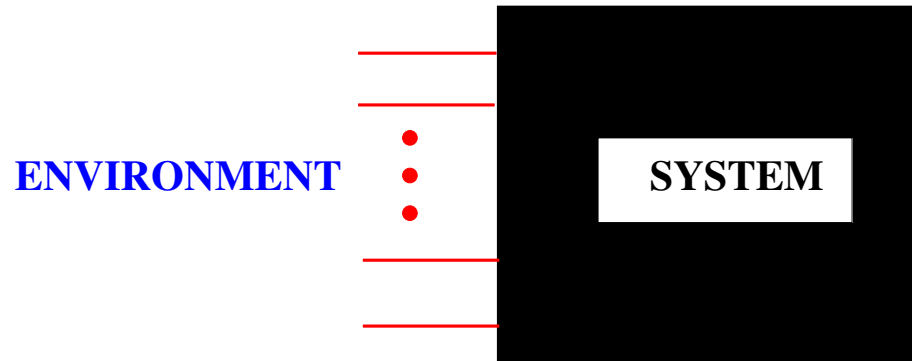
After interconnection: they are set **equal**.

Let's take a closer look at the i/o framework ...

for modeling

i/o in modeling

Physical systems often interact with their environment through **physical** terminals

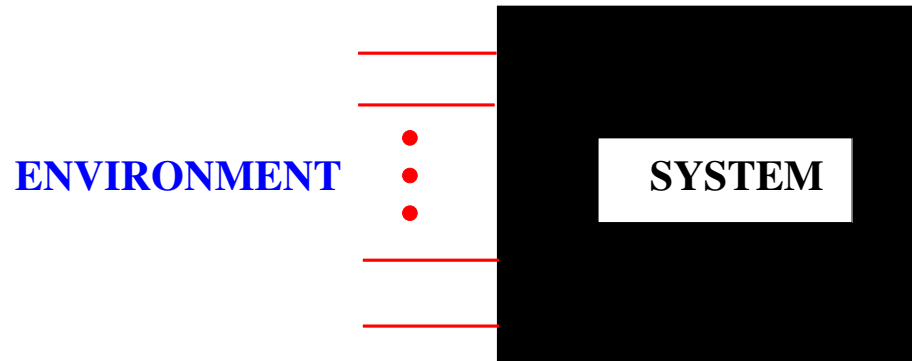


On each of these terminals many variables 'live':

- voltage & current
- position & force
- pressure & flow
- price & demand
- angle & momentum
- etc. & etc.

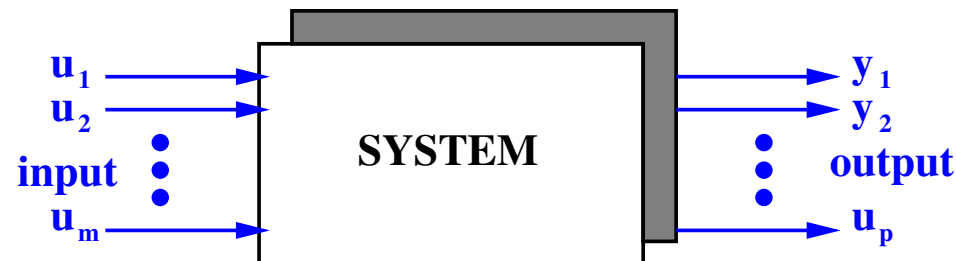
i/o in modeling

Physical systems often interact with their environment through **physical** terminals



Situation is NOT:

on one terminal there is an input, on another there is an output.



This picture is misleading, if superficially interpreted.

i/o in modeling

Physical systems often interact with their environment through **physical** terminals

The selection of what is an input and what is an output

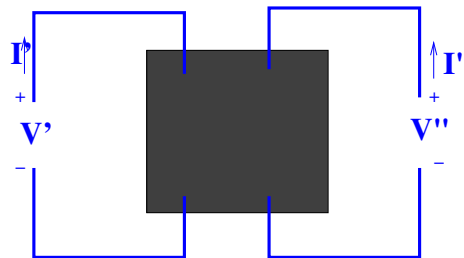
- most often does not need to be made
- if it made, it should be made **after** the modeling is done
- sometimes it **cannot** be made

i/o in modeling

Physical systems often interact with their environment through **physical** terminals

The selection of what is an input and what is an output

- does not need to be made
- if it made, it should be made **after** the modeling is done



voltage controlled?

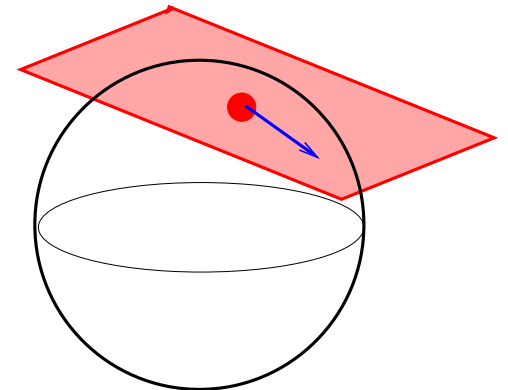
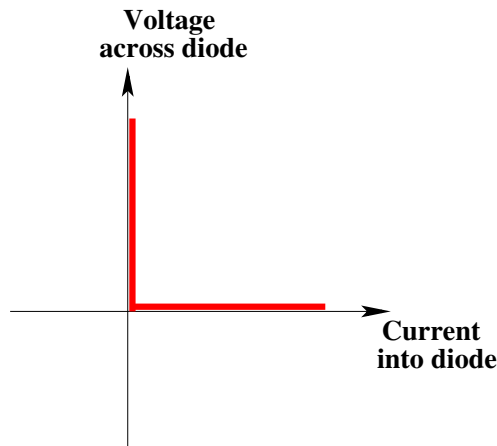
- sometimes it **cannot** be made

i/o in modeling

Physical systems often interact with their environment through **physical** terminals

The selection of what is an input and what is an output

- does not need to be made
- if it made, it should be made **after** the modeling is done
- sometimes it **cannot** be made



variables: (x, v) $\frac{d}{dt}x = v$

tangent bundle of the sphere is not 'trivial'

Conclusion

The inability of the i/o framework to properly deal with

(i) **interconnections**

and

(ii) **passive control**

is lethal.

Just as the state, the input/output partition needs to be **constructed** from first principles models. Contrary to the state, such a partition **may not be useful**, or even possible

We need a better, more flexible, universal, simpler framework that properly deals with

open & connected.

At last, a framework that deals with these difficulties

General formalism

Generalities

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

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 this universum.

This subset is what we consider the mathematical model.

Generalities

This way we arrive at the

Definition

A *math. 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.

For example, **the ideal gas law** states that the temperature T , pressure P , volume V , and quantity (number of moles) N of an ideal gas satisfy

$$\frac{PV}{NT} = R$$

with R a universal constant.

Generalities

So, before Boyle, Charles, and Avogadro got into the act, T , P , V and N may have seemed unrelated, yielding

$$\mathcal{U} = \mathbb{R}_+^4.$$

The ideal gas law restricts the possibilities to

$$\mathcal{B} = \{(T, P, V, N) \in \mathbb{R}_+^4 \mid PV/NT = R\}$$

Features

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

We will only consider **deterministic** models.

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

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

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

$P(\mathfrak{B}) =$ 'the degree of certainty (belief, plausibility, propensity, relative frequency) that outcomes are in \mathfrak{B} ;
 \cong the degree of validity of \mathfrak{B} as a model.'

We will only consider **deterministic** models.

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

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

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

Fuzzy models: there is a map μ (the 'membership function')

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

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

We will only consider **deterministic** models.

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

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

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

Determinism: $\mathcal{A} = \{\emptyset, \mathfrak{B}, \mathfrak{B}^{\text{complement}}, \mathcal{U}\}, P(\mathfrak{B}) = 1.$

Fuzzy models: there is a map μ (the 'membership function')

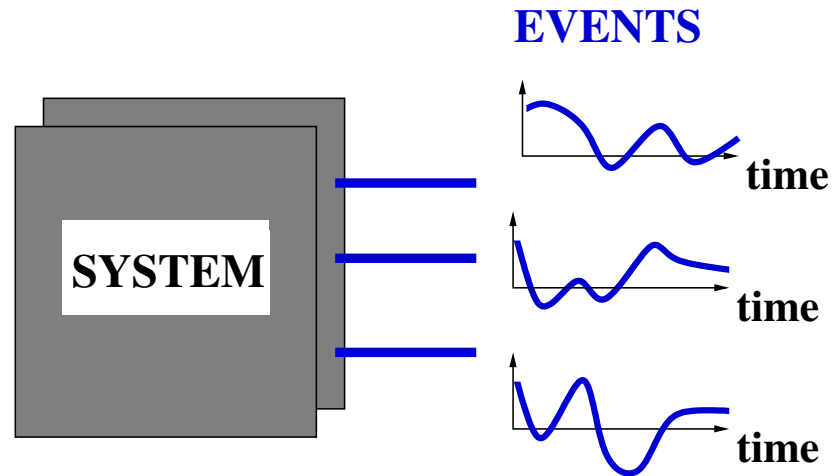
$$\mu : \mathcal{U} \rightarrow [0, 1]$$

Determinism: μ 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

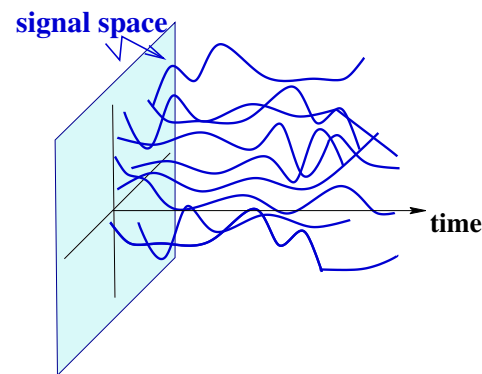
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Totality of 'legal' trajectories =: the behavior

Dynamical systems

Definition

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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 .

Dynamical systems

Definition

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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).

Dynamical systems

Definition

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

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 \mathbb{W} , the *signal space*

(= where the variables take on their values),

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

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

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).

Emphasis:

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

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

\mathcal{B} = solution set of system of (linear constant coefficient)
ODE's, or difference eqn's, or PDE's. \rightsquigarrow 'differential systems'.

A series of examples

Examples

Let's put Kepler and Newton in this setting.

K1+K2+K3 obviously define a dynamical system $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$$\mathbb{T} = \mathbb{R}, \quad \mathbb{W} = \mathbb{R}^3,$$

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

Nice example of a dynamical model 'without equations'.

Examples

Let's put Kepler and Newton in this setting.

$K_1+K_2+K_3$ obviously define a dynamical system $\Sigma = (\mathbb{T}, \mathbb{W}, \mathcal{B})$

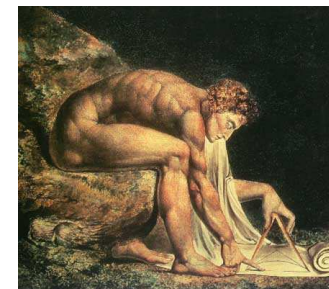
$\mathbb{T} = \mathbb{R}$, $\mathbb{W} = \mathbb{R}^3$,

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

Nice example of a dynamical model 'without equations'.

Is it a differential system?

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



Examples

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

\mathcal{B} = all state trajectories.

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

\mathcal{B} = all possible output trajectories.

Note:

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

Examples

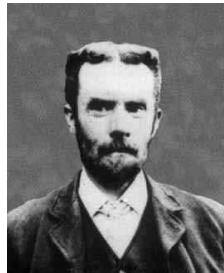
Input / output systems

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

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

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

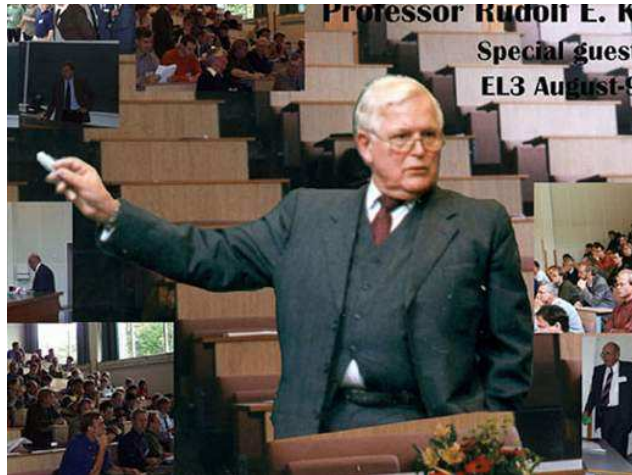
\mathfrak{B} = all input / output pairs.



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?

Examples

Codes

$\mathcal{C} \subseteq \mathbb{A}^{\mathbb{I}} = \text{the code};$ 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.**

Formal languages

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

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

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

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

Examples: All words appearing in the *Webster* dictionary

All \LaTeX documents.

Examples

Thermodynamics: a theory of **open** systems

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

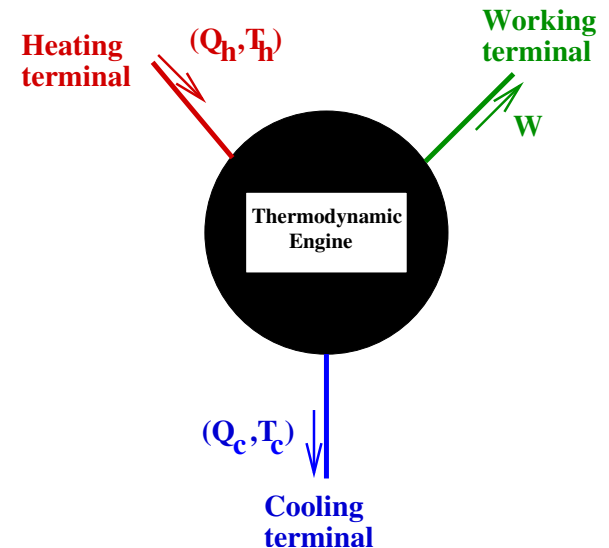
Albert Einstein

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

Arthur Eddington

Examples

Thermodynamics: a theory of **open** systems



time-axis: \mathbb{R}

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

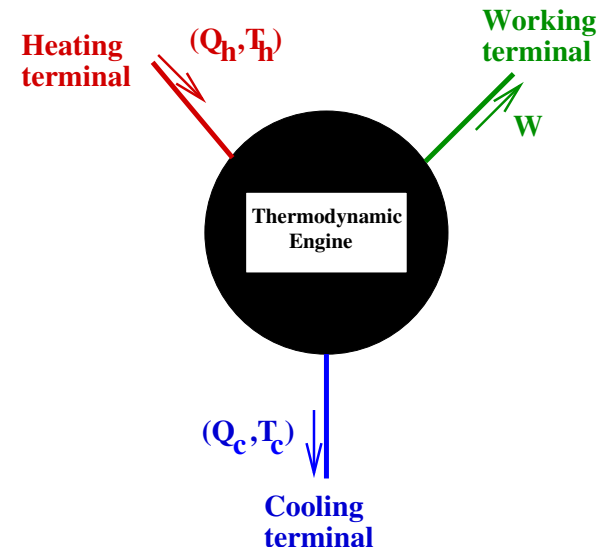
\rightsquigarrow signal space: $\mathbb{W} = \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}_+ \times \mathbb{R}$

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

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

Examples

Thermodynamics: a theory of **open** systems



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!

\mathcal{L}^\bullet : **Linear time-invariant differential systems**

More structure

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

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

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

σ^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.

or equivalently(!), completeness, or equivalently(!)

\mathfrak{B} is closed - topology of pointwise conv.

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.

$R \in \mathbb{R}^{\bullet \times w} [\xi]$ $R\left(\frac{d}{dt}\right)w = 0$ defines the
linear, time-invariant, differential system: $\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathcal{B})$ with

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

\mathcal{L}^w

$R \in \mathbb{R}^{\bullet \times w} [\xi]$ $R\left(\frac{d}{dt}\right)w = 0$ defines the
linear, time-invariant, differential system: $\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathcal{B})$ with

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

NOTATION

\mathcal{L}^\bullet : all such systems (with any - finite - number of variables)

\mathcal{L}^w : with w variables

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

$R \in \mathbb{R}^{\bullet \times w} [\xi]$ $R\left(\frac{d}{dt}\right)w = 0$ defines the
linear, time-invariant, differential system: $\Sigma = (\mathbb{R}, \mathbb{R}^w, \mathcal{B})$ with

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

NOMENCLATURE

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

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

Overview

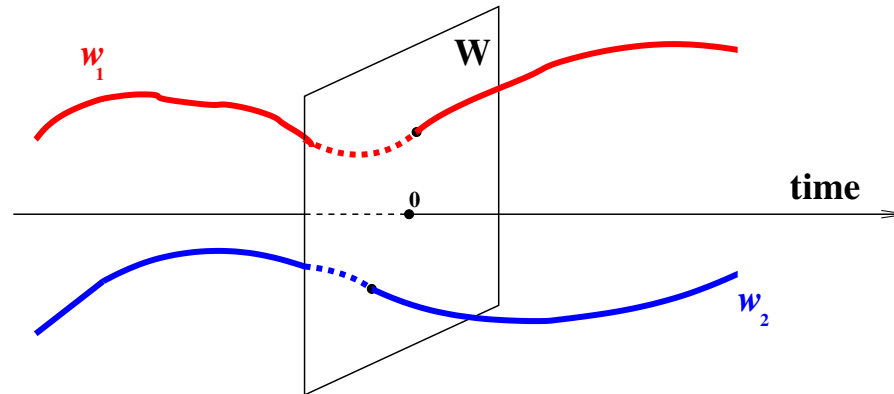
Starting from this vantage point, a rich theory has been developed

1. Modeling by **tearing**, **zooming**, and **linking**
2. **Controllability** and **stabilizability**
3. **Control by interconnection:**
from stabilization to LQ and \mathcal{H}_∞ -control
4. **Observability, observers** and the like
5. **SYSID**, the MPUM, subspace ID
6. **System representations**
7. PDE's
8. etc., etc., ...

Controllability

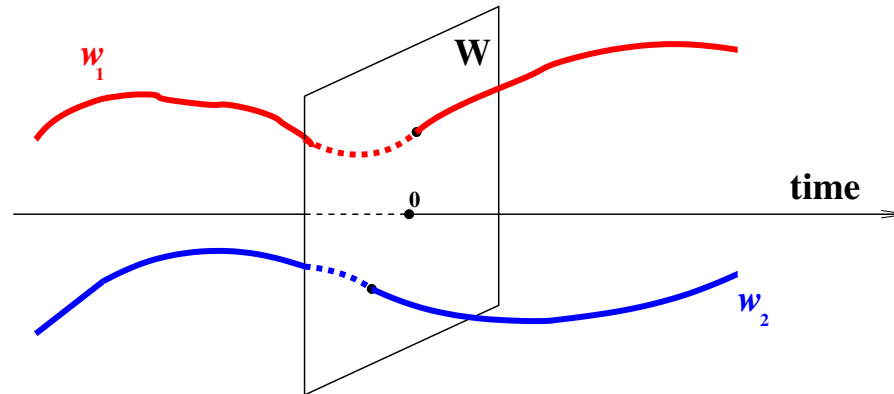
Controllability

Take any two trajectories $w_1, w_2 \in \mathcal{B}$.

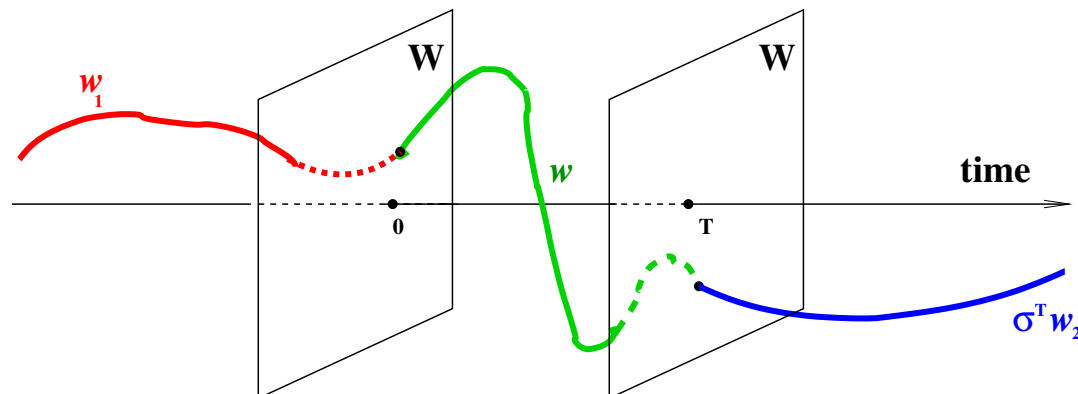


Controllability

Take any two trajectories $w_1, w_2 \in \mathcal{B}$.



Controllability:



Controllability

The time-invariant system $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$ is said to be

controllable

if for all $w_1, w_2 \in \mathfrak{B}$ there exists $w \in \mathfrak{B}$ and $T \geq 0$ such that

$$w(t) = \begin{cases} w_1(t) & t < 0 \\ w_2(t - T) & t \geq T \end{cases}$$

Controllability $:\Leftrightarrow$

legal trajectories must be **'patch-able', 'concatenable'**.

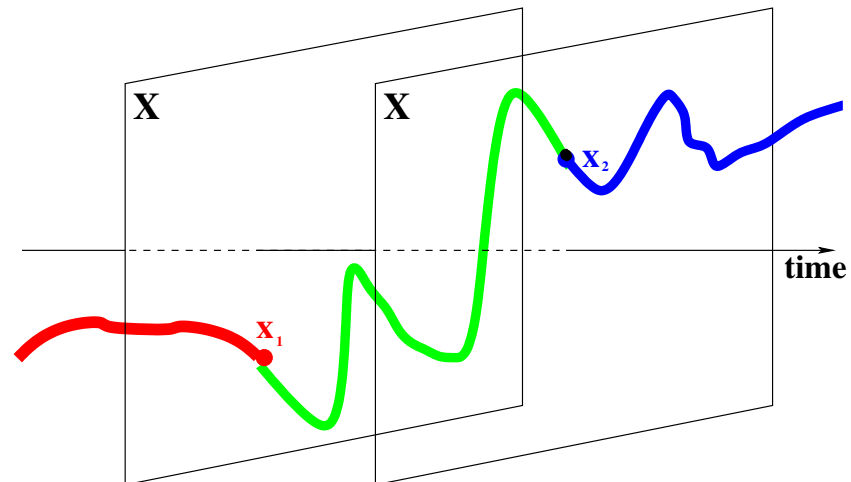
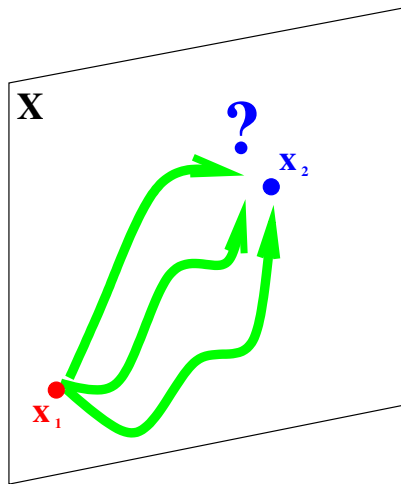
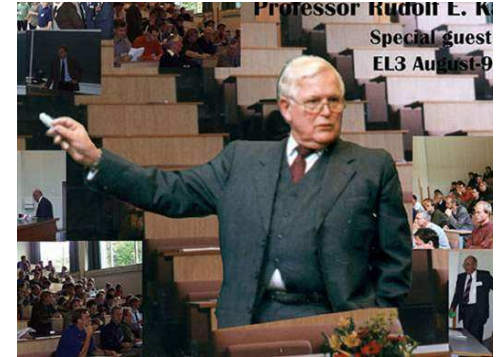
State Controllability

Special case: classical Kalman definitions for

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

controllability: variables = state or (input, state)

This is a **special case** of our controllability:



State Controllability

Special case: classical Kalman definitions for

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

controllability: variables = state or (input, state)

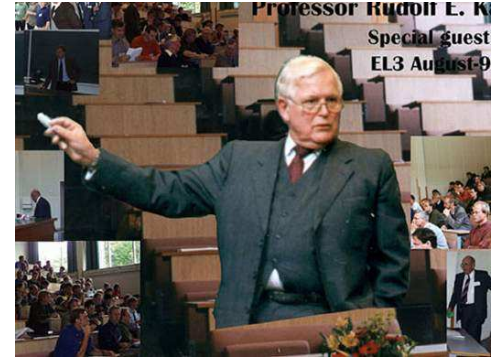
If a system is not (state) controllable, why is it?

Insufficient influence of the control?

Or bad choice of the state?

Or not properly editing the equations?

Kalman's definition addresses a rather special situation.



Tests

Given a system representation, derive algorithms in terms of the parameters for controllability.

Consider the system $\mathfrak{B} \in \mathfrak{L}^\bullet$ defined by

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

Under what conditions on $R \in \mathbb{R}^{\bullet \times w} [\xi]$ does it define a **controllable system**?

Theorem: $R \left(\frac{d}{dt} \right) w = 0$ defines a controllable system
 \Leftrightarrow
 $\text{rank} (R (\lambda)) = \text{constant over } \lambda \in \mathbb{C}.$

Tests

Notes:

• If $R \left(\frac{d}{dt} \right) w = 0$ has R of full row rank, then

controllability $\Leftrightarrow R(\lambda)$ is of full row rank $\forall \lambda \in \mathbb{C}$.

Equivalently, R is **right-invertible** as a polynomial matrix

(\Leftrightarrow **'left prime'**).

Tests

Notes:

- $\frac{d}{dt}x = Ax + Bu, w = x$ or (x, u) is controllable iff

$$\text{rank}([A - \lambda I \quad B]) = \dim(x) \quad \forall \lambda \in \mathbb{C}.$$

Popov-Belevich-Hautus test for controllability.

Of course,

$$\Leftrightarrow \text{rank} \left(\begin{bmatrix} B & AB & \dots & A^{\dim(x)-1} B \end{bmatrix} \right) = \dim(x).$$

Tests

Notes:

● When is

$$p \left(\frac{d}{dt} \right) w_1 = q \left(\frac{d}{dt} \right) w_2$$

controllable? $p, q \in \mathbb{R} [\xi]$, not both zero.

$$\text{Controllable} \iff \text{rank}([p(\lambda) \quad -q(\lambda)]) = 1 \quad \forall \lambda \in \mathbb{C}.$$

Iff p and q are co-prime. No common factors!

Testable via Sylvester matrix, etc.

Generalizable.

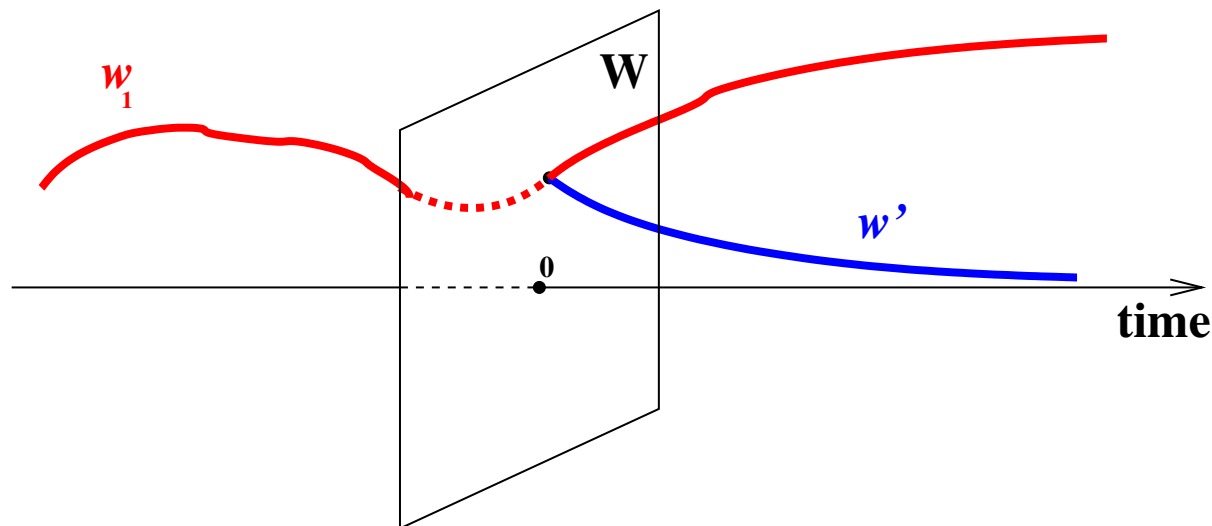
Stabilizability

The system $\Sigma = (\mathbb{T}, \mathbb{R}^w, \mathcal{B})$ is said to be **stabilizable** if, for all $w \in \mathcal{B}$, there exists $w' \in \mathcal{B}$ such that

$$w(t) = w'(t) \text{ for } t < 0 \text{ and } w'(t) \xrightarrow[t \rightarrow \infty]{} 0.$$

Stabilizability $:\Leftrightarrow$

legal trajectories can be steered to a desired point.



Stabilizability

Consider the system defined by

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

Under which conditions on $R \in \mathbb{R}^{\bullet \times w} [\xi]$ does it define a **stabilizable system**?

Theorem: $R \left(\frac{d}{dt} \right) w = 0$ defines a stabilizable system
 \Leftrightarrow
 $\text{rank} (R (\lambda)) = \text{constant over } \{\lambda \in \mathbb{C} \mid \text{Real} (\lambda) \geq 0\}.$

Image representations

Representations of \mathcal{L}^\bullet : $R \left(\frac{d}{dt} \right) w = 0$

called a *'kernel' representation*. Sol'n set $\in \mathcal{L}^\bullet$, by definition.

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

called a *'latent variable' representation* of the behavior of the w -variables.

'Elimination th'm' $\Rightarrow \in \mathcal{L}^\bullet$.

Image representations

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called a *'latent variable' representation* of the behavior of the w -variables.

'Elimination th'm' $\Rightarrow \in \mathcal{L}^\bullet$.

Missing link:

$$w = M \left(\frac{d}{dt} \right) \ell$$

called an *'image' representation* of $\mathcal{B} = \text{im} \left(M \left(\frac{d}{dt} \right) \right)$.

Elimination theorem \Rightarrow every image is also a kernel.

?? Which kernels are also images ??

Controllability!

Image representations

Theorem: (Controllability and image representations):

The following are equivalent for $\mathfrak{B} \in \mathcal{L}^\bullet$:

1. \mathfrak{B} is **controllable**
2. \mathfrak{B} admits an **image representation**

$$w = M \left(\frac{d}{dt} \right) \ell$$

3. etc., etc.

Numerical test

- **Image representation leads to an effective numerical test.**
- **∃ similar results & algorithms for time-varying systems.**
- **∃ partial results for nonlinear systems.**

Controllable part

The '**controllable part**' of $\mathfrak{B} \in \mathcal{L}^\bullet$ can be defined in many equivalent ways. Most expedient:

$$\mathfrak{B}_{\text{controllable}} := \text{largest controllable } \mathfrak{B}' \in \mathcal{L}^w, \mathfrak{B}' \subseteq \mathfrak{B}$$

Two systems

$$P_1\left(\frac{d}{dt}\right)w_1 = Q_1\left(\frac{d}{dt}\right)w_2 \quad P_2\left(\frac{d}{dt}\right)w_1 = Q_2\left(\frac{d}{dt}\right)w_2$$

have the same controllable part iff they have the same transfer function

$$P_1^{-1}Q_1 =: G_1 = G_2 := P_2^{-1}Q_2$$

Transfer function: determines the controllable part only.

Limited description. Limitation of tf. f'n manipulations.

Polynomial representations

Representations with $\mathbb{R}[\xi]$ -matrices of $\mathfrak{B} \in \mathfrak{L}^\bullet$

1. $R \left(\frac{d}{dt} \right) w = 0$ by definition
2. WLOG: R full row rank, in which case uniqueness up to pre-multiplication by unimodular
3. R left prime over $\mathbb{R}[\xi]$ ($\exists S : RS = I$) $\Leftrightarrow \mathfrak{B}$ controllable
4. $w = M \left(\frac{d}{dt} \right) \ell \Leftrightarrow \mathfrak{B}$ controllable
5. if controllable,
WLOG: M right prime over $\mathbb{R}[\xi]$ ($\exists N : NM = I$)
'observable image representation': $\exists N : \ell = N \left(\frac{d}{dt} \right) w$.

Representations with rational symbols

Let $G \in \mathbb{R}^{\bullet \times w}(\xi)$.

What does $G\left(\frac{d}{dt}\right)w = 0$ mean?

Representations with rational symbols

Let $G \in \mathbb{R}^{\bullet \times w}(\xi)$.

What does $G\left(\frac{d}{dt}\right)w = 0$ mean?

Joint work with



Yutaka Yamamoto

Representations with rational symbols

The behavior defined by $G\left(\frac{d}{dt}\right)w = 0$ is defined as that of

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

with $G = P^{-1}Q$ a left co-prime factorization over $\mathbb{R}[\xi]$ of G .

Equivalently, the output nulling behavior of

$$\frac{d}{dt}x = Ax + Bw, 0 = Cx + Dw$$

(A, B) contr., (A, C) obs., tf. f'n G .

Representations with rational symbols

The behavior defined by $G\left(\frac{d}{dt}\right)w = 0$ is defined as that of

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

with $G = P^{-1}Q$ a left co-prime factorization over $\mathbb{R}[\xi]$ of G .

Representations with $\mathbb{R}(\xi)$ -matrices of $\mathfrak{B} \in \mathcal{L}^\bullet$.

1. WLOG, with G (strictly) proper, etc.
2. G left prime over ring of stable rational f'ns $\Leftrightarrow \mathfrak{B}$ stabilizable
3. $w = G\left(\frac{d}{dt}\right)\ell \Leftrightarrow \mathfrak{B}$ controllable
4. if controllable, WLOG: G right prime over stable rational f'ns
'observable im. repr'on': $\exists F$ stable rational : $\ell = F\left(\frac{d}{dt}\right)w$.

Kucera-Youla parametrization

YK parametrization

Kucera and Youla asked the question:

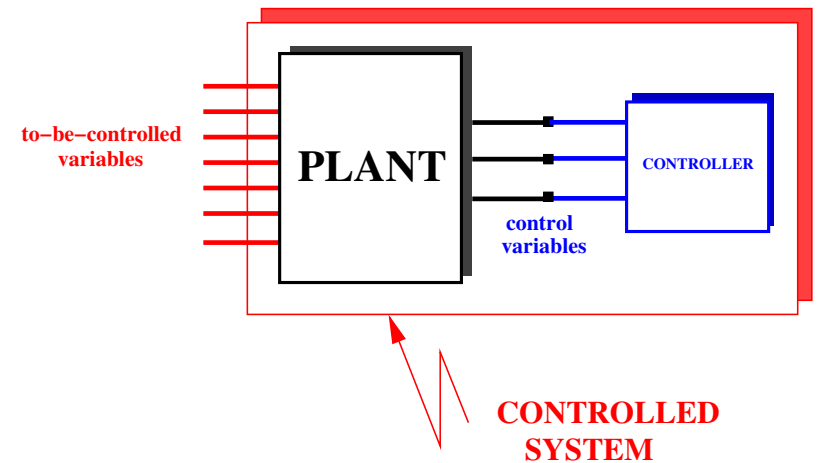
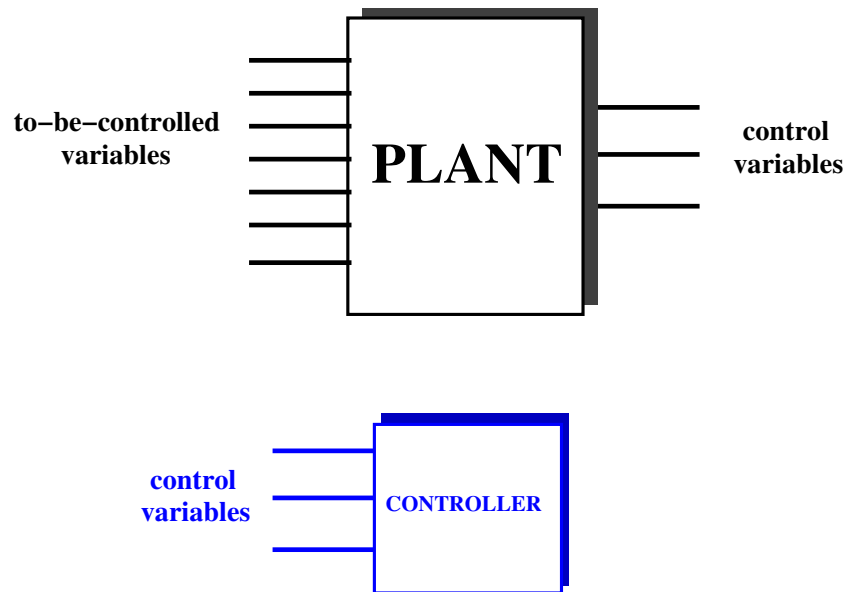
Given a system, parametrize all the stabilizing controllers. How do we deal with this question?

YK parametrization

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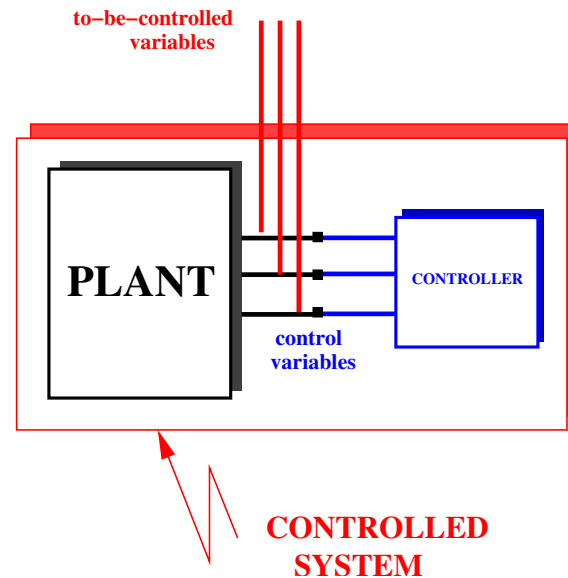
Given a system, parametrize all the stabilizing controllers. How do we deal with this question?

Control=Interconnection



YK parametrization

Control = Interconnection . We only consider ‘full control’.



$$\text{Plant: } R \left(\frac{d}{dt} \right) w = 0 \quad \text{Controller: } C \left(\frac{d}{dt} \right) w = 0$$

Given R , parametrize all C 's that lead to stability.

Stability:= ‘Hurwitz stability’ : $\Leftrightarrow w(t) \rightarrow 0$ for $t \rightarrow \infty$.

YK parametrization

Given R , parametrize all C such that $\begin{bmatrix} R \\ C \end{bmatrix}$ induces, via kernel repr., a stable system.

Given $\mathfrak{B} \in \mathcal{L}^w$, parametrize $\mathcal{K} \in \mathcal{L}^w$, such that $\mathfrak{B} \cap \mathcal{K}$ is stable.

YK parametrization

Given R , parametrize all C such that $\begin{bmatrix} R \\ C \end{bmatrix}$ induces, via kernel repr., a stable system.

Given $\mathfrak{B} \in \mathcal{L}^w$, parametrize $\mathfrak{K} \in \mathcal{L}^w$, such that $\mathfrak{B} \cap \mathfrak{K}$ is stable.

Assume \mathfrak{B} controllable. Known: If (and only if) \mathfrak{B} is controllable, $\exists \mathfrak{B}' \in \mathcal{L}^w$ such that $\mathfrak{B} \oplus \mathfrak{B}' = \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w)$.

This \mathfrak{B}' is, evidently, also controllable.

What does this say in terms of (f.r.r.) kernel repr. $R(\frac{d}{dt})w = 0$ and

$R'(\frac{d}{dt})w = 0$ of \mathfrak{B} and \mathfrak{B}' ? Obviously: $\begin{bmatrix} R \\ R' \end{bmatrix}$ unimodular.

YK parametrization

Given R , parametrize all C such that $\begin{bmatrix} R \\ C \end{bmatrix}$ induces, via kernel repr., a stable system.

Any C can therefore be written as

$$C = \begin{bmatrix} F & D \end{bmatrix} \begin{bmatrix} R \\ R' \end{bmatrix} \quad \text{i.e.} \quad C = FR + DR'$$

When does this C stabilize? Controlled system:

$$\begin{bmatrix} I & 0 \\ 0 & D \end{bmatrix} \begin{bmatrix} R \\ R' \end{bmatrix} \left(\frac{d}{dt}\right)w = 0.$$

Stable $\Leftrightarrow D$ is Hurwitz (assume D square).

YK parametrization

Given R , parametrize all C such that $\begin{bmatrix} R \\ C \end{bmatrix}$ induces, via kernel repr., a stable system.

Parametrization through $\mathbb{R}[\xi]$:

- Given R , controllable. Compute R' such that $\begin{bmatrix} R \\ R' \end{bmatrix}$ is unimodular.
- C stabilizes $\Leftrightarrow C = FR + DR'$, F free, D Hurwitz.

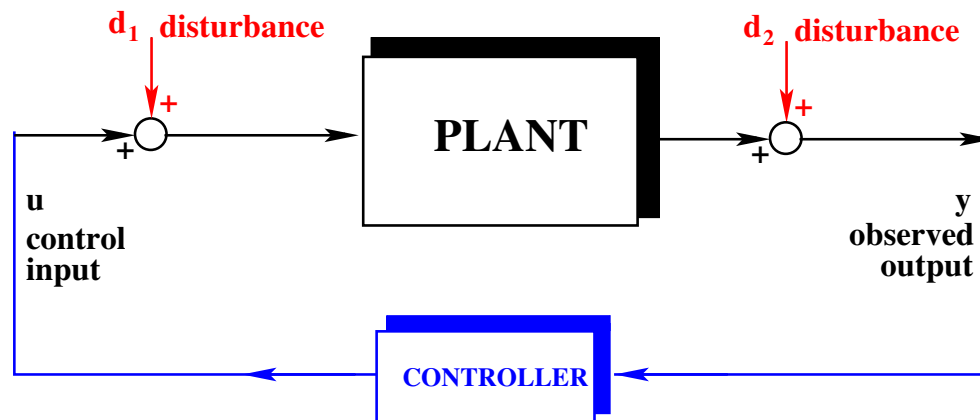
YK parametrization

YK with rational functions

$$\text{Plant: } G\left(\frac{d}{dt}\right)w = 0 \quad \text{Controller: } C\left(\frac{d}{dt}\right)w = 0$$

G, C rational. Given G , parametrize all C 's that yield stability.

Stability= 'internal stability':= all tf f'ns of additive noise structure 'stable' (in particular **proper**).



Btw, without the noise this means Hurwitz + 'regular' interconnection (cfr. Trentelman).

YK parametrization

Parametrization with polynomials:

- R controllable. R' such that $\begin{bmatrix} R \\ R' \end{bmatrix}$ is unimodular.
- C stabilizes $\Leftrightarrow C = FR + DR'$, F free, D Hurwitz.

Parametrization with rational functions:

- \mathcal{B} , stabilizable. $G(\frac{d}{dt})w = 0$ stable rational left prime repr.
- Compute G' stable rational such that $\begin{bmatrix} G \\ G' \end{bmatrix}$ is unimodular over the stable rational functions.
- C stabilizes $\Leftrightarrow C = FG + DG'$,
with F, D stable rational.

Note: nicer over stable rational: parametrization involves only stable rational ring !

PDE's

PDE's

Much of the theory also holds for PDE's.

$T = \mathbb{R}^n$, the set of independent variables, often $n = 4$,

$W = \mathbb{R}^w$, the set of dependent variables,

$\mathcal{B} =$ **sol'ns of a linear constant coefficient system of PDE's.**

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Much of the theory also holds for PDE's.

$\mathbb{T} = \mathbb{R}^n$, the set of independent variables, often $n = 4$,

$\mathbb{W} = \mathbb{R}^w$, the set of dependent variables,

$\mathcal{B} =$ sol'ns of a linear constant coefficient system of PDE's.

Let $R \in \mathbb{R}^{\bullet \times w}[\xi_1, \dots, \xi_n]$, and consider

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

Define the associated behavior

$$\mathcal{B} = \{ w \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^w) \mid (*) \text{ holds} \}.$$

Notation for n -D linear differential systems:

$$(\mathbb{R}^n, \mathbb{R}^w, \mathcal{B}) \in \mathcal{L}_n^w, \quad \text{or } \mathcal{B} \in \mathcal{L}_n^w.$$

Example

Maxwell's eq'ns, diffusion eq'n, wave eq'n, . . .



$$\begin{aligned}\nabla \cdot \vec{E} &= \frac{1}{\epsilon_0} \rho, \\ \nabla \times \vec{E} &= -\frac{\partial}{\partial t} \vec{B}, \\ \nabla \cdot \vec{B} &= 0, \\ c^2 \nabla \times \vec{B} &= \frac{1}{\epsilon_0} \vec{j} + \frac{\partial}{\partial t} \vec{E}.\end{aligned}$$

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$\mathbb{T} = \mathbb{R} \times \mathbb{R}^3$ (time and space) $n = 4$,

$$w = (\vec{E}, \vec{B}, \vec{j}, \rho)$$

(electric field, magnetic field, current density, charge density),

$$\mathbb{W} = \mathbb{R}^3 \times \mathbb{R}^3 \times \mathbb{R}^3 \times \mathbb{R}, w = 10,$$

\mathcal{B} = set of solutions to these PDE's.

Note: 10 variables, 8 equations! $\Rightarrow \exists$ free variables. 'open' system.

Submodule theorem

$R \in \mathbb{R}^{\bullet \times \bullet}[\xi_1, \dots, \xi_n]$ defines $\mathfrak{B} = \ker \left(R \left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right) \right)$,

but not vice-versa.

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

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

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

$$\mathfrak{N}_{\mathfrak{B}} := \left\{ n \in \mathbb{R}^w[\xi_1, \dots, \xi_n] \mid n^\top \left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right) \mathfrak{B} = 0 \right\}.$$

Proposition:

$\mathfrak{N}_{\mathfrak{B}}$ is a $\mathbb{R}[\xi_1, \dots, \xi_n]$ sub-module of $\mathbb{R}^w[\xi_1, \dots, \xi_n]$.

Submodule theorem

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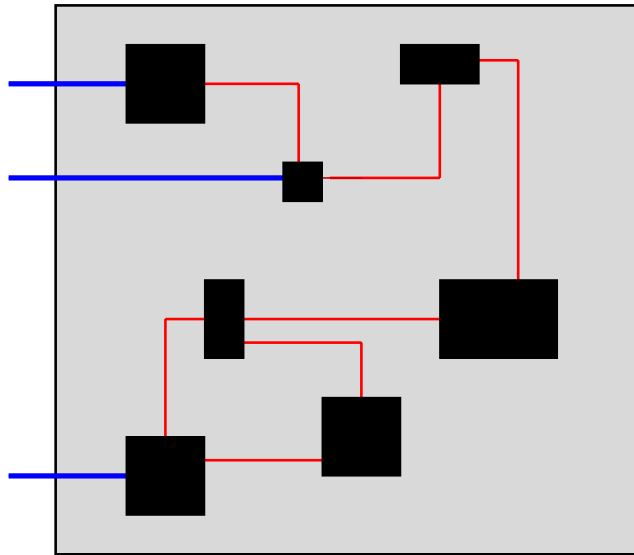
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Theorem:

$\mathfrak{L}_n^w \xleftrightarrow{\text{bijective}} \text{submodules of } \mathbb{R}^w[\xi_1, \dots, \xi_n]$

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).

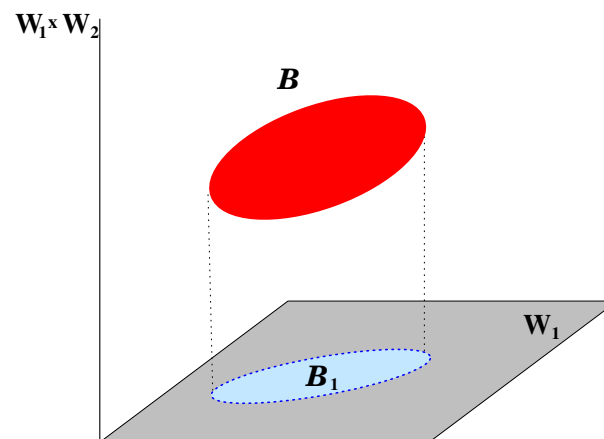
¿ Can these latent variables be eliminated from the equations ?

Elimination theorem

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

$$R_1\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right)w_1 = R_2\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right)w_2.$$

Obviously, the behavior of the (w_1, w_2) 's is described by a system of PDE's. **¿ Is the behavior of the w_1 's alone also ?**



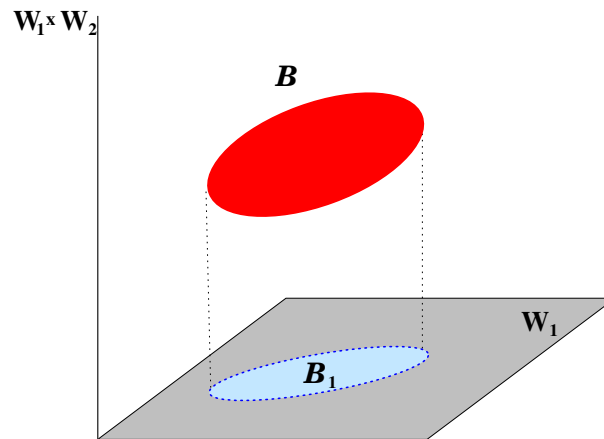
Elimination theorem

In the language of behaviors:

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

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

Does this 'projection' \mathcal{B}_1 belong to $\mathcal{L}_n^{w_1}$?



Theorem:

It does!

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

The Fundamental Principle

Proof: 'Fundamental principle'. Consider

$$F(x) = y$$

Given: $F : X \rightarrow Y$, $y \in Y$; Unknown: $x \in X$.

¿ Does there exist a sol'n x ?

Examples:

- 1.
- 2.
- 3.

The Fundamental Principle

Proof: 'Fundamental principle'. Consider

$$F(\boldsymbol{x}) = \boldsymbol{y}$$

Given: $F : X \rightarrow Y$, $\boldsymbol{y} \in Y$; Unknown: $\boldsymbol{x} \in X$.

¿ Does there exist a sol'n \boldsymbol{x} ?

Examples:

1. $F \in \mathbb{R}^{n_1 \times n_2}$, $\boldsymbol{y} \in \mathbb{R}^{n_2}$, $\boldsymbol{x} \in \mathbb{R}^{n_1}$

2.

3.

The Fundamental Principle

Proof: 'Fundamental principle'. Consider

$$F(\boldsymbol{x}) = \boldsymbol{y}$$

Given: $F : \mathbb{X} \rightarrow \mathbb{Y}$, $\boldsymbol{y} \in \mathbb{Y}$; Unknown: $\boldsymbol{x} \in \mathbb{X}$.

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Examples:

1.

2. ODE's:

$$F\left(\frac{d}{dt}\right)\boldsymbol{x} = \boldsymbol{y}$$

with $F \in \mathbb{R}^{n_1 \times n_2}[\xi]$, $\boldsymbol{y} \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^{n_2})$, $\boldsymbol{x} \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^{n_1})$.

Or over distributions, $\boldsymbol{y} \in \mathcal{D}'(\mathbb{R}, \mathbb{R}^{n_2})$, $\boldsymbol{x} \in \mathcal{D}'(\mathbb{R}, \mathbb{R}^{n_1})$.

The Fundamental Principle

Proof: 'Fundamental principle'. Consider

$$F(\boldsymbol{x}) = \boldsymbol{y}$$

Given: $F : \mathbb{X} \rightarrow \mathbb{Y}$, $\boldsymbol{y} \in \mathbb{Y}$; Unknown: $\boldsymbol{x} \in \mathbb{X}$.

¿ Does there exist a sol'n \boldsymbol{x} ?

Examples:

1.

2.

3. PDE's:

$$F\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right)\boldsymbol{x} = \boldsymbol{y}$$

$F \in \mathbb{R}^{n_1 \times n_2}[\xi_1, \dots, \xi_n]$, $\boldsymbol{y} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_2})$,

$\boldsymbol{x} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_1})$, or over distributions.

The Fundamental Principle for PDE's

$$F\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right) \mathbf{x} = \mathbf{y}$$

Given: $F \in \mathbb{R}^{n_1 \times n_2}[\xi_1, \dots, \xi_n]$, $\mathbf{y} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_2})$,

Unknown: $\mathbf{x} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_1})$.

¿ Does there exist a sol'n \mathbf{x} ?

Obvious necessary condition:

$$\begin{aligned} (n \in \mathbb{R}^{n_1}[\xi_1, \dots, \xi_n]) \wedge (n^\top(\xi_1, \dots, \xi_n)F(\xi_1, \dots, \xi_n) = 0) \\ \Rightarrow n^\top\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right)\mathbf{y} = 0. \end{aligned}$$

Theorem (Fundamental principle): This is a n.a.s.c.

The Fundamental Principle for PDE's

$$F\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right) \mathbf{x} = \mathbf{y}$$

Given: $F \in \mathbb{R}^{n_1 \times n_2}[\xi_1, \dots, \xi_n]$, $\mathbf{y} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_2})$,

Unknown: $\mathbf{x} \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}^{n_1})$.

¿ Does there exist a sol'n \mathbf{x} ?

Theorem (Fundamental principle): This is a n.a.s.c.

Since the n 's form a (finitely generated) $\mathbb{R}[\xi_1, \dots, \xi_n]$ -module, this is a finite condition!

Example:

Take $0 \neq F \in \mathbb{R}[\xi_1, \dots, \xi_n]$. PDE $F\left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n}\right) \mathbf{x} = \mathbf{y}$.

Always solvable!

The elimination theorem

There exist effective algorithms for $(R_1, R_2) \mapsto R$.

↪ Computer algebra, Gröbner bases.

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

- **Intersection:** $(\mathcal{B}_1, \mathcal{B}_2 \in \mathcal{L}_n^W) \Rightarrow (\mathcal{B}_1 \cap \mathcal{B}_2 \in \mathcal{L}_n^W)$
- **Addition:** $(\mathcal{B}_1, \mathcal{B}_2 \in \mathcal{L}_n^W) \Rightarrow (\mathcal{B}_1 + \mathcal{B}_2 \in \mathcal{L}_n^W)$
- **Projection:** $(\mathcal{B} \in \mathcal{L}_n^{w_1+w_2}) \Rightarrow (\Pi_{w_1} \mathcal{B} \in \mathcal{L}_n^{w_1}) \Pi_{w_1}$: projection
- **Action of a linear differential operator:**
 $(\mathcal{B} \in \mathcal{L}_n^{w_1}, P \in \mathbb{R}^{w_2 \times w_1}[\xi_1, \dots, \xi_n]) \Rightarrow (P(\frac{d}{dt})\mathcal{B} \in \mathcal{L}_n^{w_2}).$
- **Inverse image of a linear differential operator:**
 $(\mathcal{B} \in \mathcal{L}_n^{w_2}, P \in \mathbb{R}^{w_2 \times w_1}[\xi_1, \dots, \xi_n]) \Rightarrow (P(\frac{d}{dt}))^{-1}\mathcal{B} \in \mathcal{L}_n^{w_1}.$

Elimination theorem

Which PDE's describe (ρ, \vec{E}, \vec{j}) in Maxwell's equations ?

Eliminate \vec{B} from Maxwell's equations \rightsquigarrow

$$\begin{aligned}\nabla \cdot \vec{E} &= \frac{1}{\epsilon_0} \rho, \\ \epsilon_0 \frac{\partial}{\partial t} \nabla \cdot \vec{E} + \nabla \cdot \vec{j} &= 0, \\ \epsilon_0 \frac{\partial^2}{\partial t^2} \vec{E} + \epsilon_0 c^2 \nabla \times \nabla \times \vec{E} + \frac{\partial}{\partial t} \vec{j} &= 0.\end{aligned}$$

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

is called a **kernel representation** of the associated $\mathfrak{B} \in \mathfrak{L}_n^w$.

Another representation: **image representation**

$$w = M \left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right) \ell.$$

‘Elimination’ thm $\Rightarrow \text{im} \left(M \left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right) \right) \in \mathfrak{L}_n^w !$

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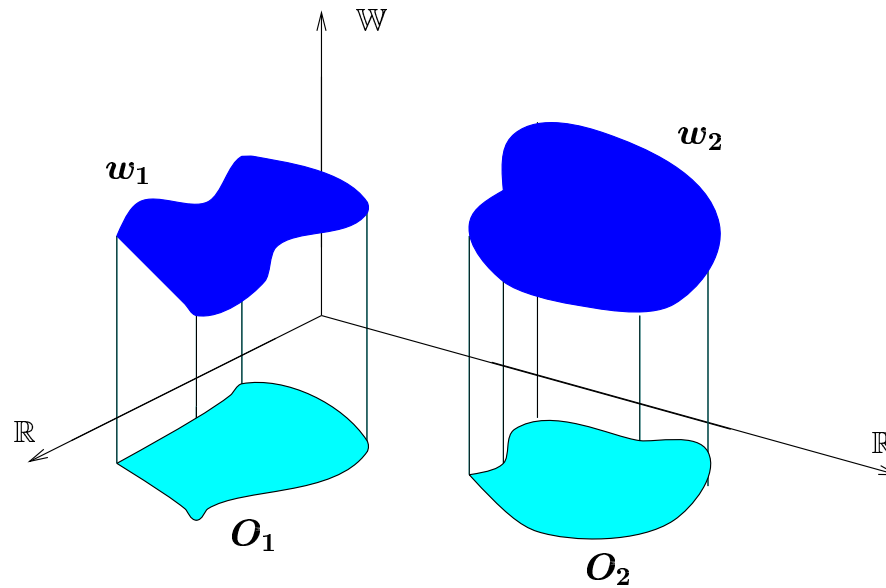
‘Elimination’ thm $\Rightarrow \text{im} \left(M \left(\frac{\partial}{\partial x_1}, \dots, \frac{\partial}{\partial x_n} \right) \right) \in \mathfrak{L}_n^w !$

Which linear diff. systems admit an image representation???

$\mathfrak{B} \in \mathfrak{L}_n^w$ admits an image representation iff it is **‘controllable’**.

Controllability for PDE's

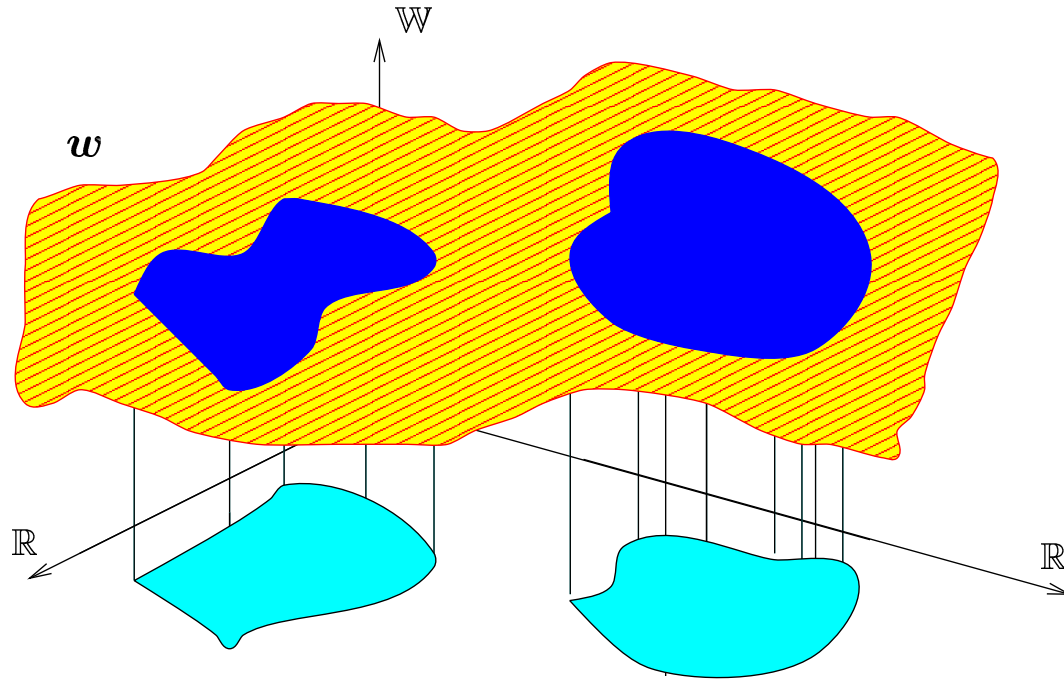
Controllability def'n in pictures:



$$w_1, w_2 \in \mathcal{B}.$$

Controllability for PDE's

$\exists w \in \mathfrak{B}$ 'patches' $w_1, w_2 \in \mathfrak{B}$.



Controllability \Leftrightarrow 'patch-ability'.

Are Maxwell's equations controllable ?

The following equations in the *scalar potential* $\phi : \mathbb{R} \times \mathbb{R}^3 \rightarrow \mathbb{R}$ and the *vector potential* $\vec{A} : \mathbb{R} \times \mathbb{R}^3 \rightarrow \mathbb{R}^3$, generate exactly the solutions to Maxwell's equations:

$$\vec{E} = -\frac{\partial}{\partial t}\vec{A} - \nabla\phi,$$

$$\vec{B} = \nabla \times \vec{A},$$

$$\vec{j} = \epsilon_0 \frac{\partial^2}{\partial t^2} \vec{A} - \epsilon_0 c^2 \nabla^2 \vec{A} + \epsilon_0 c^2 \nabla (\nabla \cdot \vec{A}) + \epsilon_0 \frac{\partial}{\partial t} \nabla \phi,$$

$$\rho = -\epsilon_0 \frac{\partial}{\partial t} \nabla \cdot \vec{A} - \epsilon_0 \nabla^2 \phi.$$

Proves controllability. Illustrates the interesting connection

controllability $\Leftrightarrow \exists$ potential!

Conclusion

The flexibility and generality of the behavioral approach in modeling, for system representations, for passive control, dealing with PDE's, etc. is evident.

Exemplified by the notion of controllability.

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**Nature and Nature's laws lay hid in night
God said, 'Let Newton be' and all was light**

Conclusion

The flexibility and generality of the behavioral approach in modeling, for system representations, for passive control, dealing with PDE's, etc. is evident.

Exemplified by the notion of controllability.

**Nature and Nature's laws lay hid in night
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**Mathematical Systems Theory lay bound by might
Ratio said, 'Let Behaviors be' and all was right**

Details & copies of the lecture frames are available from/at

Jan.Willems@esat.kuleuven.be

<http://www.esat.kuleuven.be/~jwillems>

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Thank you

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