



Extracting the memory of a system:

STATE CONSTRUCTION

Chaire Francqui, Lecture III, May 12, 2004

UCL Université catholique de Louvain

STATE SPACE SYSTEMS

THEME

How do we formalize the **memory** of a dynamical system?

When is a variable a **state variable**?

How do state equations look like?

How are state equations constructed, algorithmically, starting from other representations ?

To what extent are state representations of a system **unique**?

The state is a latent variable with special properties ...

... *What's a latent variable system ?*

LATENT VARIABLES

‘First principles’ models contain **auxiliary** variables:

- interconnection variables in modeling
- state variables is systems theory
- the nodes in trellis diagrams in coding
- the labels of the vertices in automata
- non-terminal symbols in formal languages
- Ω is probability theory

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- Ω is probability theory

We now first formalize dynamical systems which contain such variables, called **latent** variables,

↔ **manifest** variables (whose behavior the model aims at).

LATENT VARIABLES

A dynamical system with latent variables =

$$\Sigma_{\mathbb{L}} = (\mathbb{T}, \mathbb{W}, \mathbb{L}, \mathcal{B}_{\text{full}})$$

$\mathbb{T} \subset \mathbb{R}$, the *time-axis* (= the set of relevant time instances)

\mathbb{W} , the *signal space* (= space of variables the model aims at)

\mathbb{L} , the *latent variable space* (= space of **auxiliary** model variables)

$$\mathcal{B}_{\text{full}} \subset (\mathbb{W} \times \mathbb{L})^{\mathbb{T}} : \text{the full behavior}$$

(= the pairs $(w, l) : \mathbb{T} \rightarrow \mathbb{W} \times \mathbb{L}$ the model declares possible)

THE MANIFEST BEHAVIOR

Call the elements of \mathbb{W} *'manifest' variables* ,
those of \mathbb{L} *'latent' variables* .

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The latent variable system $\Sigma_L = (\mathbb{T}, \mathbb{W}, \mathbb{L}, \mathcal{B}_{\text{full}})$ induces
the *manifest system* $\Sigma = (\mathbb{T}, \mathbb{W}, \mathcal{B})$, with *manifest behavior*

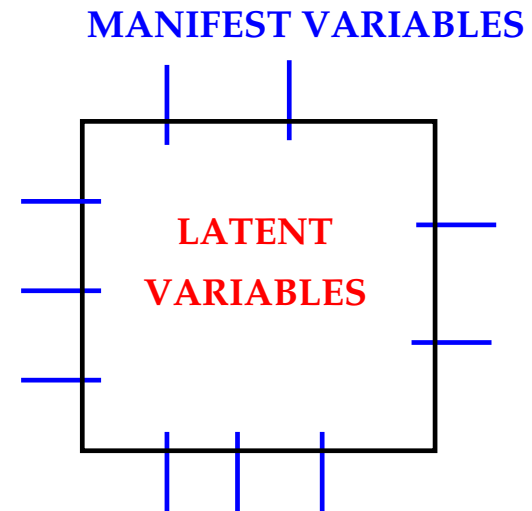
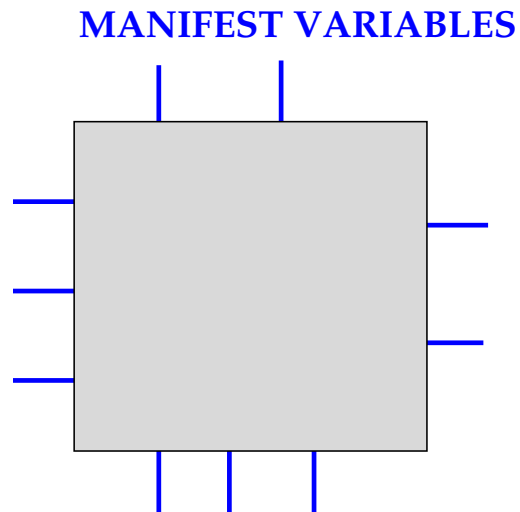
$$\mathcal{B} = \{w : \mathbb{T} \rightarrow \mathbb{W} \mid \exists l : \mathbb{T} \rightarrow \mathbb{L} \text{ such that } (w, l) \in \mathcal{B}_{\text{full}}\}$$

In equations for \mathcal{B} , the latent variables are *'eliminated'*.

Note the clear notion of **equivalent models** that emerges!

THE MANIFEST BEHAVIOR

Call the elements of \mathbb{W} *'manifest' variables* ,
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THE NOTION OF STATE

A state system :=

A latent variable system in which the latent variable has a special ‘splitting’ property.

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A state system :=

A **latent variable** system in which the latent variable has a special '**splitting**' property.

The **latent variable system**

$$\Sigma_X = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{\text{full}})$$

is said to be a **state system** if

$$(\mathbf{w}_1, \mathbf{x}_1), (\mathbf{w}_2, \mathbf{x}_2) \in \mathfrak{B}_{\text{full}}, t_0 \in \mathbb{T}, \text{ and } \mathbf{x}_1(t_0) = \mathbf{x}_2(t_0)$$

imply

$$(\mathbf{w}_1, \mathbf{x}_1) \underset{t_0}{\wedge} (\mathbf{w}_2, \mathbf{x}_2) \in \mathfrak{B}_{\text{full}}.$$

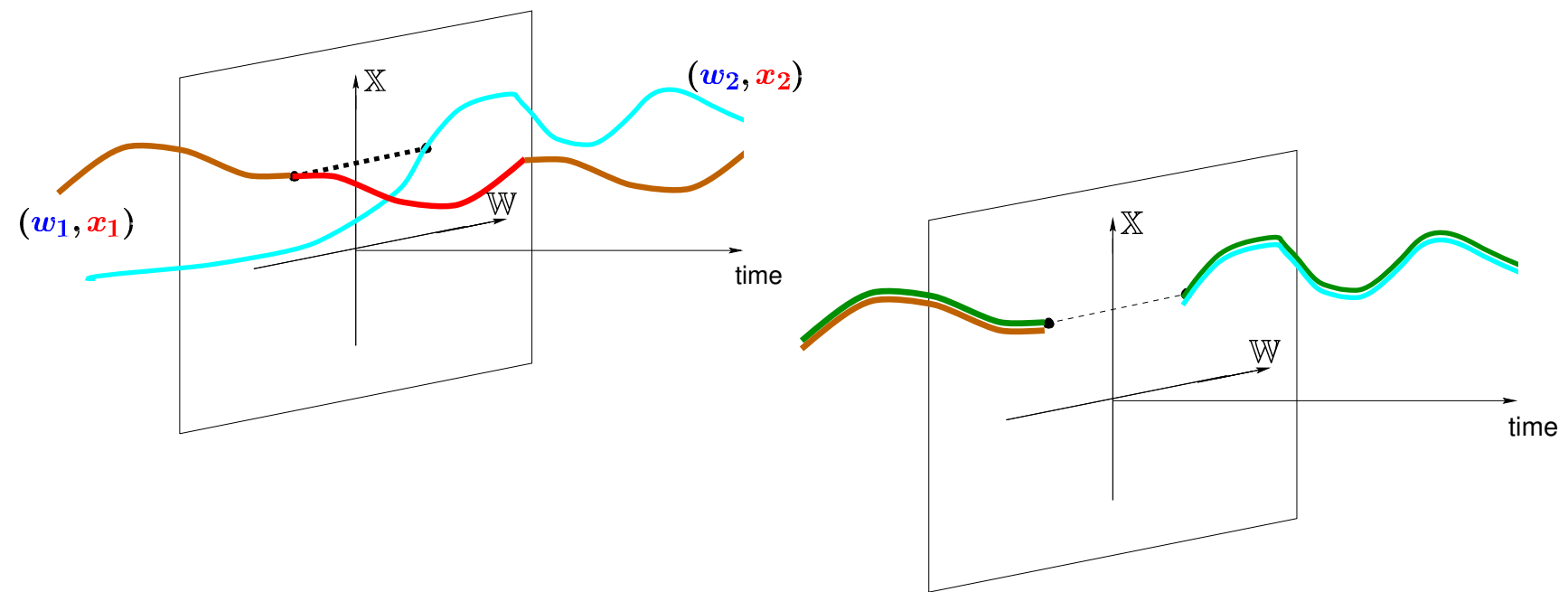
\wedge_{t_0} denotes *concatenation* at t_0 , defined as

$$f_1 \wedge_{t_0} f_2(t) := \begin{cases} f_1(t) & \text{for } t < t_0 \\ f_2(t) & \text{for } t \geq t_0 \end{cases}$$

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In pictures:



This definition is the implementation of the idea:

The state at time t , $\mathbf{x}(t)$, contains all the information (about (\mathbf{w}, \mathbf{x}) !) that is relevant for the future behavior.

The state = the **memory**.

The **past** and the **future** are 'independent', conditioned on (given) the **present state**.

\cong Markovianity! Splitting!

Examples of state systems:

1. Discrete-time systems.

A latent variable system described by a difference equation that is *first order* in the **latent** variable x , and *zero-th order* in the **manifest** variable w :

$$F(x(t+1), x(t), w(t), t) = 0.$$

Examples of state systems:

1. Discrete-time systems.

2. Continuous-time systems.

A latent variable system described by a differential equation that is *first order* in the **latent** variable x , and *zero-th order* in the **manifest** variable w :

$$F\left(\frac{d}{dt}x(t), x(t), w(t), t\right) = 0.$$

In particular, the ubiquitous

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

Examples of state systems:

1. Discrete-time systems.
2. Continuous-time systems.
3. Automata.
4. Trellis diagrams.
5. QM: $\frac{d}{dt}\psi = i\hbar H(\psi)$, $p = |\psi|^2$;

ψ = the ‘wave function’;

$p(x, t)$ = the ‘probability’ density of the particle’s position.

wave function = latent, state, observables = manifest ??

For discrete time state systems \rightsquigarrow

Theorem: The latent variable system

$$\Sigma_X = (\mathbb{Z}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{\text{full}})$$

is a state system

if (and only if,
provided the behavior is **'complete'**) $\mathfrak{B}_{\text{full}}$ admits a representation
as a difference equation that is

*first order in the latent variable x , and
zero-th order in the manifest variable w :*

$$F(x(t+1), x(t), w(t), t) = 0.$$

STATE FOR DIFFERENTIAL SYSTEMS

Here we meet the notorious \mathcal{C}^∞ -difficulty:

concatenation and \mathcal{C}^∞ don't mix.

We hence modify the state axiom to:

The **latent variable system** $\Sigma_X = (\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathfrak{B}_{\text{full}}) \in \mathcal{L}^{w+n}$ is said to be a **state system** if

$$(\boldsymbol{w}_1, \boldsymbol{x}_1), (\boldsymbol{w}_2, \boldsymbol{x}_2) \in \mathfrak{B}_{\text{full}}, t_0 \in \mathbb{T}, \quad \text{and} \quad \boldsymbol{x}_1(t_0) = \boldsymbol{x}_2(t_0)$$

$$\text{imply} \quad (\boldsymbol{w}_1, \boldsymbol{x}_1) \underset{t_0}{\wedge} (\boldsymbol{w}_2, \boldsymbol{x}_2) \in \mathfrak{B}_{\text{full}}^{\text{closure}}.$$

'Closure' w.r.t., the \mathcal{L}^{loc} -topology.

\Leftrightarrow : if $(\boldsymbol{w}_1, \boldsymbol{x}_1) \underset{t_0}{\wedge} (\boldsymbol{w}_2, \boldsymbol{x}_2)$ is a **weak sol'n** of corr. ODE.

DESCRIPTOR SYSTEMS

Theorem: The latent variable system $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}_{\text{full}}) \in \mathcal{L}^{w+n}$ is a state system

if and only if

$\mathcal{B}_{\text{full}}$ admits a kernel representation that is **first order in the latent variable x** , and **zero-th order in the manifest variable w** .

In other words, iff there exist $E, F, G \in \mathbb{R}^{\bullet \times \bullet}$ such that this kernel representation takes the form of a **descriptor system**:

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

MINIMALITY

We can consider two types of **minimality of state representations**:

1. Minimality of **the number of equations**
2. Minimality of **the number of state variables**

We discuss mainly the second one.

Definition: The state system $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}_{\text{full}}) \in \mathcal{L}^{w+n}$ is said to be **state-minimal** if, whenever $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^{n'}, \mathcal{B}'_{\text{full}}) \in \mathcal{L}^{w+n'}$ is another state system with the same manifest behavior, there holds

$$n \leq n'.$$

TRIMNESS

One more definition...

$\mathcal{B} \in \mathcal{L}^w$ is said to be **trim** if, $\forall w_0 \in \mathbb{R}^w, \exists w \in \mathcal{B}$ such that $w(0) = w_0$. The state system $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}_{\text{full}}) \in \mathcal{L}^{w+n}$ is said to be **state-trim** if, $\forall x_0 \in \mathbb{R}^n, \exists (w, x) \in \mathcal{B}_{\text{full}}$ such that $x(0) = x_0$.

Theorem: (Minimality of state representations):

The state system $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}_{\text{full}}) \in \mathcal{L}^{w+n}$ is **state-minimal** iff it is **state trim** and the state x is **observable from w** .

Observability : $\Leftrightarrow x$ can be deduced from w .

I.e., $\exists X \in \mathbb{R}^{n \times w}[\xi]$ such that

$$(w, x) \in \mathcal{B}_{\text{full}} \Rightarrow x = X\left(\frac{d}{dt}\right)w.$$

FURTHER RESULTS

1. State isomorphism theorem.

Assume $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}_{\text{full}})$ $(\mathbb{R}, \mathbb{R}^w, \mathbb{R}^n, \mathcal{B}'_{\text{full}}) \in \mathcal{L}^{w+n}$, both state-minimal, same manifest behavior

\Rightarrow there exists a nonsingular $S \in \mathbb{R}^{n \times n}$ such that

$$[(w, x) \in \mathcal{B}_{\text{full}} \text{ and } (w, x') \in \mathcal{B}'_{\text{full}}] \Leftrightarrow [x' = Sx].$$

The minimal state representation is unique up to a choice of the basis in the state space.

FURTHER RESULTS

1. State isomorphism theorem.

2. Controllability.

The manifest behavior is **controllable** iff there exists a state representation of it whose state behavior is controllable.

If the manifest behavior is **controllable** then any state-minimal state representation of it is **state-controllable**.

FURTHER RESULTS

1. State isomorphism theorem.

2. Controllability.

3. Descriptor systems.

\exists algorithms acting on E, F, G in a descriptor repr'ion to verify its state-minimality, its eq'n minimality, both.

$$E \frac{d}{dt} x + Fx + Gw = 0 \quad \text{and} \quad E' \frac{d}{dt} x' + F'x' + G'w = 0$$

are two minimal (**state- & eq'ion-minimal**) repr'ions of the same man. b'ior iff \exists nonsingular $T, S \in \mathbb{R}^{\bullet \times \bullet}$ such that

$$E' = TES, F' = TES, G' = TG.$$

Uniqueness up to choice of basis in state and equation space.

FURTHER RESULTS

1. State isomorphism theorem.
2. Controllability.
3. Descriptor systems.
4. Notation: Introduce the ‘invariant’ (the ‘state cardinality’)

$$n : \mathcal{L}^\bullet \rightarrow \mathbb{N}$$

$n(\mathcal{B})$:= dimension of the **minimal** state associated with \mathcal{B} .

All 'classical' results remain valid, except, (fortunately!)
the celebrated (non-)equivalence:
state-minimality \Leftrightarrow state-observability + state-controllability.

Non-controllable systems are **very 'real'** and they allow
state-minimal (non-controllable) state representation.

I/S/O SYSTEMS

Finally...

It is possible to combine the **input/output partition and the state representation**, leading to the ubiquitous:

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

\mathbf{u} is input := free,

\mathbf{y} is output := bound by \mathbf{u} ,

\mathbf{x} is state := 'splitting' = memory.

Notation:

$$\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$$

Theorem: Let $\mathfrak{B} \in \mathcal{L}^\bullet$.

There exists a componentwise partition $w = (u, y)$, with $\dim(u) = m(\mathfrak{B})$, $\dim(y) = p(\mathfrak{B})$, and matrices

$$A \in \mathbb{R}^{n(\mathfrak{B}) \times n(\mathfrak{B})}, B \in \mathbb{R}^{n(\mathfrak{B}) \times m(\mathfrak{B})}, C \in \mathbb{R}^{p(\mathfrak{B}) \times n(\mathfrak{B})}, D \in \mathbb{R}^{p(\mathfrak{B}) \times m(\mathfrak{B})}$$

such that

$$\frac{d}{dt}x = Ax + Bu, \quad y = Cx + Du,$$

is a minimal (eq'n- and state-minimal) state repr'ion of \mathfrak{B} .

Note important 'invariants' ('cardinalities):

$$w(\mathfrak{B}), m(\mathfrak{B}), p(\mathfrak{B}), n(\mathfrak{B})$$

associated with $\mathfrak{B} \in \mathcal{L}^\bullet$.

$\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$ is minimal (state + eq'n minimal)

\Leftrightarrow it is state-minimal

\Leftrightarrow it is state-observable

$$\Leftrightarrow \text{rank} \left(\begin{bmatrix} C \\ CA \\ \vdots \\ CA^{\dim(A)-1} \end{bmatrix} \right) = \dim(A).$$

$\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$ is state controllable (usual Kalman def'n)

$$\Leftrightarrow \text{rank}([B \ AB \ \dots \ A^{\dim(A)-1}B]) = \dim(A).$$

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If $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$ is minimal (i.e., observable) then

state controllable **iff** manifest behavior controllable.

Watch out:

minimality of $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] \Leftarrow$ **but** $\not\Rightarrow$ controllable & observable.

STATE CONSTRUCTION

**!! Given a dynamical system $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$
find a state representation $\Sigma_X = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{\text{full}})$
for it !!**

This problem is a jewel that has emerged in systems theory (and in computer science) in the 1960's. It has ramifications in the theory of stochastic processes (where they hide under the name **hidden Markov models - HMM's**), in formal language theory, (more recently) model simplification, etc.

See my webpage for the general set theoretic construction, using 'Nerode equivalence'.

We only consider linear time-invariant differential (c.q. difference) systems.

STATE CONSTRUCTION in DIFFERENTIAL SYSTEMS

!! Given a representation of the manifest behavior

$$\mathfrak{B} \in \mathcal{L}^\bullet,$$

find a (state-minimal) state representation for it !!

Problem:

Given a ‘numerical’ specification of the (manifest) behavior, end up with a ‘numerical’ specification of a state model.

ALGORITHMS

- Given the **impulse response** construct a state model $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$.

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↪ the theory around the **Hankel matrix**.

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- Given **data** (e.g. $u(\cdot), u(\cdot)$), construct a state model $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$.
- Make sure $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$ is in a special (e.g., **balanced**) form

STATE MAPS

Let $X(\xi) \in \mathbb{R}^{\bullet \times w}[\xi]$. The map $X\left(\frac{d}{dt}\right)$ is called a **state map** for $\mathfrak{B} \in \mathcal{L}^w$ if the full behavior

$$\mathfrak{B}_{\text{full}} = \left\{ (w, x) \mid w \in \mathfrak{B} \text{ and } x = X\left(\frac{d}{dt}\right)w \right\}$$

satisfies the axiom of state. **Minimal state map:** obvious.

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In a state-minimal repr'on, x is determined by a state map **(because of observability)**, whence (minimal) state maps exist.

State representation problem: Find a minimal state map.

Most logical : **latent variable** repr'on \rightsquigarrow state repr'on.
However, we only discuss **kernel** and **image** repr'ons.

Define the **'shift-and-cut'** operator σ on $\mathbb{R}[\xi]$ as follows:

$$\begin{aligned}\sigma : p_0 + p_1\xi + \cdots + p_{n-1}\xi^{n-1} + p_n\xi^n \\ \mapsto p_1 + p_2\xi + \cdots + p_{n-1}\xi^{n-2} + p_n\xi^{n-1}\end{aligned}$$

Extend-able in the obvious term-by-term way to $\mathbb{R}^{\bullet \times \bullet}[\xi]$.

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Repeated use of the cut-and-shift on $P \in \mathbb{R}^{\bullet \times \bullet}[\xi]$ yields the **'stack'** operator Σ_P , defined by

$$\Sigma_P := \begin{bmatrix} \sigma(P) \\ \sigma^2(P) \\ \vdots \\ \sigma^{\text{degree}(P)}(P) \end{bmatrix}$$

FROM KERNEL to STATE REPRESENTATION

There is a construction (elegant in its simplicity) of a state map in terms of the cut-and-shift and stack operators!

Theorem: Let

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

be a kernel representation of $\mathfrak{B} \in \mathcal{L}^w$.

Then $\Sigma_R\left(\frac{d}{dt}\right)$ is a state map for \mathfrak{B} . The resulting

$$R\left(\frac{d}{dt}\right)w = 0 ; \quad x = \Sigma_R\left(\frac{d}{dt}\right)w$$

need not be minimal. It is trivially state-observable, but it may not be state-trim. Using **Gröbner basis techniques** it can be trimmed, leading to a minimal state representation.

FROM IMAGE to STATE REPRESENTATION

Theorem: Let $w = M\left(\frac{d}{dt}\right)\ell$
be an image representation of $\mathfrak{B} \in \mathfrak{L}_{\text{cont}}^w$. Then $\Sigma_M\left(\frac{d}{dt}\right)$ (acting on ℓ) yields a state map for \mathfrak{B} . The resulting

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We obtain a state map that acts on ℓ . If $w = M\left(\frac{d}{dt}\right)\ell$ is not observable, then the state may not be observable, whence not state-minimal. But, if the image representation is observable, then so is the state. Hence then there is a(n implicit) state map. State trim iff the rows of Σ_M are linearly independent over \mathbb{R} . So, trimming only requires deleting rows.

The image repr. gives a very effective state construction.

SINGLE INPUT - SINGLE OUTPUT SYSTEMS

Apply this to

$$p\left(\frac{d}{dt}\right)y = q\left(\frac{d}{dt}\right)u$$

with

$$p(\xi) = p_0 + p_1\xi + \cdots + p_{n-1}\xi^{n-1} + p_n\xi^n, \quad p_n \neq 0$$

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There are 4 well-known state constructions:

1. the *observer canonical form*
2. the *observability canonical form*
3. the *controllability canonical form*
4. the *controller canonical form*

SINGLE INPUT - SINGLE OUTPUT SYSTEMS

The cut-and-shift and stack operators yield the polynomial matrix

$$\Sigma_R(\xi) = \begin{bmatrix} p_1 + \dots + p_{n-1}\xi^{n-2} + p_n\xi^{n-1} & -q_1 - \dots - q_{n-1}\xi^{n-2} - q_n\xi^{n-1} \\ p_2 + \dots + p_{n-1}\xi^{n-3} + p_n\xi^{n-2} & -q_2 - \dots - q_{n-1}\xi^{n-3} - q_n\xi^{n-2} \\ \vdots & \vdots \\ p_{n-1} + p_n\xi & -q_{n-1} - q_n\xi \\ p_n & -q_n \end{bmatrix}$$

It follows that $x = \Sigma_R\left(\frac{d}{dt}\right)$ is a state map, in fact, a **state minimal one**, even if the system is not controllable, i.e., when p and q have a **common factor**.

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To get more convenient minimal state maps, we can take any basis for span of the rows of X .

SINGLE INPUT - SINGLE OUTPUT SYSTEMS

To get more convenient minimal state maps, we can take any basis for span of the rows of X .

One choice: take the rows of Σ_R in reverse order.

A small calculation shows that this choice of the state variables leads to the so-called **observer canonical form**, the i/s/o representation

$$A = \begin{bmatrix} -p_{n-1}/p_n & 1 & 0 & \cdots & 0 & 0 \\ -p_{n-2}/p_n & 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ -p_0/p_n & 0 & 0 & \cdots & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} q_{n-1} - p_{n-1}q_n/p_n \\ q_{n-2} - p_{n-2}q_n/p_n \\ \vdots \\ q_0 - p_0q_n/p_n \end{bmatrix},$$
$$C = [1/p_n \ 0 \ 0 \ \cdots \ 0 \ 0], \quad D = [q_n/p_n].$$

Another immediate choice is to pick the state map

$$X(\xi) = \begin{bmatrix} 1 & \star \\ \xi & \star \\ \vdots & \vdots \\ \xi^{n-2} & \star \\ \xi^{n-1} & \star \end{bmatrix}$$

We need to compute the \star 's so that the combinations of the rows of Σ_R that yield the first column of X also give the second column.

The second column can be obtained by long hand division of q by p , i.e., by computing the polynomial $b(\xi) \in \mathbb{R}[\xi]$ defined by the equation

$$p(\xi)b(\xi^{-1}) = q(\xi) \quad (\text{modulo } \xi^{-1}\mathbb{R}[\xi^{-1}]).$$

$$\text{Then } X(\xi) = \begin{bmatrix} 1 & & & & b_0 \\ \xi & & & & b_1 + b_0\xi \\ \vdots & & & & \vdots \\ \xi^{n-2} & & & & b_{n-2} + b_{n-3}\xi + \dots + b_0\xi^{n-2} \\ \xi^{n-1} & & & & b_{n-1} + b_{n-2}\xi + \dots + b_0\xi^{n-1} \end{bmatrix} \cdot$$

This leads to the **observable canonical form**, the i/s/o representation

$$A = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ -\frac{p_0}{p_n} & -\frac{p_1}{p_n} & -\frac{p_2}{p_n} & \dots & -\frac{p_{n-1}}{p_n} \end{bmatrix}, \quad B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_{n-1} \\ b_n \end{bmatrix},$$

$$C = [1 \ 0 \ \dots \ 0 \ 0], \quad D = [b_0].$$

SINGLE INPUT - SINGLE OUTPUT SYSTEMS

When the SISO system is controllable, and given in image representation by

$$\begin{bmatrix} u \\ y \end{bmatrix} = \begin{bmatrix} p\left(\frac{d}{dt}\right) \\ q\left(\frac{d}{dt}\right) \end{bmatrix} \ell$$

with

$$p(\xi) = p_0 + p_1\xi + \cdots + p_{n-1}\xi^{n-1} + p_n\xi^n, \quad p_n \neq 0,$$
$$q(\xi) = q_0 + q_1\xi + \cdots + q_{n-1}\xi^{n-1} + q_n\xi^n.$$

SINGLE INPUT - SINGLE OUTPUT SYSTEMS

The cut-and-shift and stack operators yield

$$X(\xi) = \begin{bmatrix} p_1 + \cdots + p_{n-1} \xi^{n-2} + p_n \xi^{n-1} \\ q_1 + \cdots + q_{n-1} \xi^{n-2} + q_n \xi^{n-1} \\ p_2 + \cdots + p_{n-1} \xi^{n-3} + p_n \xi^{n-2} \\ q_2 + \cdots + q_{n-1} \xi^{n-3} + q_n \xi^{n-2} \\ \vdots \\ p_{n-1} + p_n \xi \\ q_{n-1} + q_n \xi \\ q_n \\ p_n \end{bmatrix} \cdot$$

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There are again two ready bases for the linear span of the rows of X :

$$\begin{bmatrix} p_n \\ p_{n-1} + p_n \xi \\ \vdots \\ p_2 + \cdots + p_{n-1} \xi^{n-3} + p_n \xi^{n-2} \\ p_1 + \cdots + p_{n-1} \xi^{n-2} + p_n \xi^{n-1} \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 1 \\ \xi \\ \vdots \\ \xi^{n-2} \\ \xi^{n-1} \end{bmatrix} \cdot$$

The first choice leads to the **controllable canonical form**

$$\begin{aligned} A &= \begin{bmatrix} -p_{n-1}/p_n & 1 & 0 & \cdots & 0 & 0 \\ -p_{n-2}/p_n & 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ -p_0/p_n & 0 & 0 & \cdots & 0 & 1 \end{bmatrix}, & B &= \begin{bmatrix} 0 \\ 0 \\ \vdots \\ \vdots \\ 1 \end{bmatrix}, \\ C &= [b_1 \ b_2 \ \cdots \ b_{n-1} \ b_n], & D &= [b_0]. \end{aligned}$$

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 \end{aligned}$$

The second choice leads to the **controller canonical form**

$$\begin{aligned}
 A &= \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ -\frac{p_0}{p_n} & -\frac{p_1}{p_n} & -\frac{p_2}{p_n} & \cdots & -\frac{p_{n-1}}{p_n} \end{bmatrix}, & B &= \begin{bmatrix} 0 \\ 0 \\ \vdots \\ \frac{1}{p_n} \end{bmatrix}, \\
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$$C = [q_0 - p_0 \frac{q_n}{p_n} \quad q_1 - p_1 \frac{q_n}{p_n} \quad \cdots \quad q_{n-1} - p_{n-1} \frac{q_n}{p_n}], \quad D = [q_n].$$

Minimality (observability) holds iff p and q are co-prime.

FROM LATENT VARIABLE to STATE REPRESENTATION

Consider the latent variable system

$\Sigma_X = (\mathbb{R}, \mathbb{R}^{w_1+w_2}, \mathbb{R}^n, \mathcal{B}_{\text{full}}) \in \mathcal{L}^{w_1+w_2+n}$. Eliminate $w_2 \rightsquigarrow$
 $\Sigma'_X = (\mathbb{R}, \mathbb{R}^{w_1}, \mathbb{R}^n, \mathcal{B}'_{\text{full}})$. It is easy to deduce directly from the
state axiom that Σ'_X is a state system if Σ_X is.

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Construction of a state representation for \mathcal{B} :

1. $R(\frac{d}{dt})w = M(\frac{d}{dt})\ell$ **latent variable** representation for \mathcal{B} .
2. Apply the **cut-and-shift and stack operators** to $[R \mid -M]$.
3. Obtain a state map

$$x = \Sigma_{[R \mid -M]} \left(\frac{d}{dt} \right) \begin{bmatrix} w \\ \ell \end{bmatrix}.$$

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\rightsquigarrow a, not nec. minimal, latent var'ble state repr'ion for \mathcal{B} .

Notes

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This complements the existing algorithms

transfer function \rightarrow i / s / o representation;

impulse response \rightarrow i / s / o representation.

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- Basic idea of algorithms: **from a (e.g. latent variable) representation directly to state model.**
- \exists **Gröbner basis** algorithms for state trimming.
- Our state construction is easily extended to state / input construction.
- Examples of useful special (minimal) state repr'ons:

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i/s/o representation:

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

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output nulling representation:

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driving variable representation:

$$\frac{d}{dt}\mathbf{x} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{v}, \quad \mathbf{w} = \mathbf{C}\mathbf{x} + \mathbf{D}\mathbf{v}.$$

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Readily deduced from descriptor representation:

$$E \frac{d}{dt} \mathbf{x} + F \mathbf{x} + G \mathbf{w} = 0.$$

Notes

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- Our state construction is easily extended to state / input construction.
- Examples of useful special (minimal) state repr'ons:
- Recent advances: from image representation directly to **balanced** state representation.

End of the Lecture III