# The Behavioral Approach to Systems Theory

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# Outline

#### The axiom of state

Discrete-time systems

First-order representations

State maps

The shift-and-cut map

Algebraic characterization

Continuous-time systems

Computation of state-space representations

• Are state representations "natural"?

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  - First principles and "tearing and zooming" modelling
     → systems of high-order differential equations

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- What does that imply for the equations?
- How to construct a state from the equations?
- How to construct a state representation from the equations?

## The basic idea

It's the Mariners' final game in the World Series. You're late...



The current score is what matters...

# The basic idea

- The state contains all the relevant information about the future behavior of the system
- The state is the memory of the system
- Independence of past and future given the state

## The axiom of state

#### $\boldsymbol{\Sigma} = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{full})$ is a *state system* if

$$(w_1, x_1), (w_2, x_2) \in \mathfrak{B}_{\mathrm{full}} \text{ and } x_1(T) = x_2(T) \ 
onumber \ (w_1, x_1) \bigwedge_T (w_2, x_2) \in \mathfrak{B}_{\mathrm{full}}$$

 $\bigwedge_{T}$  is concatenation at T:

$$(f_1 \wedge f_2)(t) := \left\{ egin{array}{c} f_1(t) ext{ for } t < T \ f_2(t) ext{ for } t \geq T \end{array} 
ight.$$

## Graphically...

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# 



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## Example 1: discrete-time system

#### $\Sigma = (\mathbb{Z}, \mathbb{R}^{w}, \mathbb{R}^{1}, \mathfrak{B}_{full})$ , with

$$\mathfrak{B}_{\text{full}} := \{ (\mathbf{W}, \ell) \mid \mathbf{F} \circ (\sigma \ell, \ell, \mathbf{W}) = \mathbf{0} \}$$

where

$$\sigma: (\mathbb{R}^1)^{\mathbb{Z}} \to (\mathbb{R}^1)^{\mathbb{Z}}$$
  
 $(\sigma(\ell))(k) := \ell(k+1)$ 

## Example 1: discrete-time system

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$$\mathfrak{B}_{\mathrm{full}} := \{(\mathbf{W}, \ell) \mid \mathbf{F} \circ (\sigma \ell, \ell, \mathbf{W}) = \mathbf{0}\}$$

#### Special case: input-state-output equations

$$\sigma x = f(x, u)$$
  

$$y = h(x, u)$$
  

$$w = (u, y)$$

### Example 2: continuous-time system

 $\Sigma = (\mathbb{R}, \mathbb{R}^{w}, \mathbb{R}^{1}, \mathfrak{B}_{full})$ , with

$$\mathfrak{B}_{\mathrm{full}} := \{(w, \ell) \mid F \circ (\frac{d}{dt}\ell, \ell, w) = 0\}$$

#### Special case: input-state-output equations

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

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#### Theorem: A 'complete' latent variable system

$$\boldsymbol{\Sigma} = (\mathbb{Z}, \mathbb{R}^{\mathsf{w}}, \mathbb{R}^{\mathsf{x}}, \mathfrak{B}_{\mathrm{full}})$$

is a state system iff  $\,\mathfrak{B}_{full}$  can be described by

$$\boldsymbol{F}\circ(\boldsymbol{\sigma}\boldsymbol{x},\boldsymbol{x},\boldsymbol{w})=\boldsymbol{0}$$

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0-th order in *w*, 1st order in *x* 

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Linear case:

$$\boldsymbol{E}\boldsymbol{\sigma}\boldsymbol{x} + \boldsymbol{F}\boldsymbol{x} + \boldsymbol{G}\boldsymbol{w} = \boldsymbol{0}$$

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1st order in x is consequence of state property!

## Proof (linear case)

$$\mathcal{V} := \left\{ \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mid \exists (x, w) \in \mathfrak{B}_{\text{full}} \text{ s. t. } \begin{bmatrix} x(1) \\ x(0) \\ w(0) \end{bmatrix} = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \right\}$$

 $\mathcal{V}$  linear  $\Rightarrow \exists E, F, G \text{ s.t. } \mathcal{V} = \text{ker}( \begin{bmatrix} E & F & G \end{bmatrix})$ 

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 $\mathcal{V}$  linear  $\Rightarrow \exists E, F, G$  s.t.  $\mathcal{V} = \text{ker}( \begin{bmatrix} E & F & G \end{bmatrix})$ 

 $[(\mathbf{x}, \mathbf{w}) \in \mathfrak{B}_{\text{full}} \Longrightarrow \mathbf{E}\sigma\mathbf{x} + \mathbf{F}\mathbf{x} + \mathbf{G}\mathbf{w} = \mathbf{0}]$ 

∜

# Proof (linear case)

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 $\mathcal{V}$  linear  $\Rightarrow \exists E, F, G$  s.t.  $\mathcal{V} = \text{ker}([E F G])$ 

Converse by induction, using axiom of state:

 $E\sigma x + Fx + Gw = 0$  on  $[0, k] \Longrightarrow (w, x)_{|[0,k]} \in \mathfrak{B}_{\mathrm{full}|[0,k]}$ 

Then apply completeness of  $\mathfrak{B}$ 

State construction: basic idea

# **Problem:** Given kernel or hybrid description, find a state representation

 $\boldsymbol{E}\boldsymbol{\sigma}\boldsymbol{x} + \boldsymbol{F}\boldsymbol{x} + \boldsymbol{G}\boldsymbol{w} = \boldsymbol{0}$ 

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**First** compute polynomial operator in the shift acting on system variables, inducing a state variable:

 $X(\sigma)w = x$ 

$$X(\sigma) \left[ \begin{array}{c} \mathbf{w} \\ \ell \end{array} \right] = \mathbf{x}$$

State construction: basic idea

# **Problem:** Given kernel or hybrid description, find a state representation

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First compute polynomial operator in the shift acting on system variables, inducing a state variable:

$$X(\sigma) W = x \qquad \qquad X(\sigma) \begin{bmatrix} W \\ \ell \end{bmatrix} = x$$

Then use original eqs. and *X* to obtain 1st order representation.

State maps for kernel representations

 $X \in \mathbb{R}^{\bullet \times w}[\xi]$  induces a state map  $X(\sigma)$  for ker $(R(\sigma))$  if the behavior  $\mathfrak{B}_{full}$  with latent variable x, described by

$$\begin{array}{rcl} R(\sigma)w &=& 0\\ X(\sigma)w &=& x \end{array}$$

satisfies the axiom of state.

## Example

$$\mathfrak{B} = \{ \mathbf{w} \mid \mathbf{r}(\sigma)\mathbf{w} = \mathbf{0} \}$$

where  $r \in \mathbb{R}[\xi]$ , deg(r) = n.

(Minimal) state map induced by

$$\begin{bmatrix} 1\\ \xi\\ \vdots\\ \xi^{n-1} \end{bmatrix} \rightsquigarrow \begin{bmatrix} w\\ \sigma W\\ \vdots\\ \sigma^{n-1}W \end{bmatrix}$$

The axiom of state revisited

A *linear* system  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{full})$  with latent variable *x* is a state system if

$$(w, x) \in \mathfrak{B}_{\mathrm{full}} \mathrm{and} x(T) = 0$$
 $\downarrow \downarrow$ 
 $(\mathbf{0}, \mathbf{0}) \bigwedge_{T} (w, x) \in \mathfrak{B}_{\mathrm{full}}$ 

The axiom of state revisited

A *linear* system  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{full})$  with latent variable *x* is a state system if

• Time-invariance  $\implies$  can choose T = 0;

The axiom of state revisited

A *linear* system  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{full})$  with latent variable *x* is a state system if

- Time-invariance  $\implies$  can choose T = 0;
- Concatenability with zero trajectory is key.

## When is $w \in \mathfrak{B}$ concatenable with zero?

$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$

•••	0	0	w(0)	w(1)	w(2)	w(3)	•••
	k = -2	k = -1	k = 0	<i>k</i> = 1	<i>k</i> = 2	<i>k</i> = 3	
$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$
  
... 0 0  $R_0$   $R_1$   $R_2$   $R_3$  ...  
... 0 0 w(0) w(1) w(2) w(3) ...  
... k = -2  $k = -1$   $k = 0$   $k = 1$   $k = 2$   $k = 3$  ...

 $R_0w(0) + R_1w(1) + \ldots + R_Lw(L) = 0$ 

$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$

•••	0	<b>R</b> 0	<b>R</b> 1	<b>R</b> <sub>2</sub>	<b>R</b> 3	<b>R</b> 4	•••
•••	0	0	w(0)	w(1)	w(2)	w(3)	•••
•••	k = -2	k = -1	k = 0	<i>k</i> = 1	k = 2	<i>k</i> = 3	•••

 $R_0w(0) + R_1w(1) + \ldots + R_Lw(L) = 0$  $R_1w(0) + R_2w(1) + \ldots + R_Lw(L-1) = 0$ 

$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$

•••	$R_0$	<b>R</b> 1	<b>R</b> <sub>2</sub>	<b>R</b> 3	<b>R</b> <sub>4</sub>	<b>R</b> 5	•••
•••	0	0	w(0)	w(1)	w(2)	w(3)	
•••	k = -2	k = -1	<i>k</i> = 0	<i>k</i> = 1	<i>k</i> = 2	<i>k</i> = 3	•••

 $R_0w(0) + R_1w(1) + \ldots + R_Lw(L) = 0$   $R_1w(0) + R_2w(1) + \ldots + R_Lw(L-1) = 0$  $R_2w(0) + R_3w(1) + \ldots + R_Lw(L-2) = 0$ 

$$\mathbf{R}_0\mathbf{w} + \mathbf{R}_1\sigma\mathbf{w} + \ldots + \mathbf{R}_L\sigma^L\mathbf{w} = \mathbf{0}$$

•••	0	0	w(0)	w(1)	w(2)	w(3)	• • •
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= :

$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$

•••	<b>R</b> <sub>L-3</sub>	<b>R</b> <sub>L-2</sub>	<b>R</b> <sub>L-1</sub>	RL	0	0	•••
•••	0	0	<i>w</i> (0)	w(1)	w(2)	w(3)	•••
•••	k = -2	k = -1	k = 0	<i>k</i> = 1	<i>k</i> = 2	<i>k</i> = 3	

$$R_0 w(0) + R_1 w(1) + \ldots + R_L w(L) = 0$$
  

$$R_1 w(0) + R_2 w(1) + \ldots + R_L w(L-1) = 0$$
  

$$R_2 w(0) + R_3 w(1) + \ldots + R_L w(L-2) = 0$$
  

$$\vdots \qquad = \vdots$$
  

$$R_{L-1} w(0) + R_L w(1) = 0$$

$$R_0 w + R_1 \sigma w + \ldots + R_L \sigma^L w = 0$$
  
...  $R_{L-2} R_{L-1} R_L 0 0 0 \ldots$   
... 0 0 w(0) w(1) w(2) w(3) ...  
...  $k = -2$   $k = -1$   $k = 0$   $k = 1$   $k = 2$   $k = 3$  ...

 $R_{0}w(0) + R_{1}w(1) + \ldots + R_{L}w(L) = 0$   $R_{1}w(0) + R_{2}w(1) + \ldots + R_{L}w(L-1) = 0$   $R_{2}w(0) + R_{3}w(1) + \ldots + R_{L}w(L-2) = 0$   $\vdots \qquad = \vdots$   $R_{L-1}w(0) + R_{L}w(1) = 0$   $R_{L}w(0) = 0$ 

The shift-and-cut map

$$egin{aligned} &\sigma_+:\mathbb{R}[\xi] o\mathbb{R}[\xi]\ &\sigma_+(\sum_{i=0}^nm{p}_i\xi^i):=\sum_{i=0}^{n-1}m{p}_{i+1}\xi^i \end{aligned}$$

#### "Divide by $\xi$ and take polynomial part"

#### Extended componentwise to vectors and matrices

 $R(\xi) = R_0 + R_1 \xi + \ldots + R_{L-1} \xi^{L-1} + R_L \xi^L$ 

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 $\sigma_{+}(R(\xi)) = R_{1} + \ldots + R_{L-1}\xi^{L-2} + R_{L}\xi^{L-1}$ 

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$$\vdots = \vdots$$

$$R(\xi) = R_0 + R_1 \xi + \ldots + R_{L-1} \xi^{L-1} + R_L \xi^L$$

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$$\sigma_{+}^{2}(\mathbf{R}(\xi)) = \mathbf{R}_{2} + \ldots + \mathbf{R}_{L-1}\xi^{L-3} + \mathbf{R}_{L}\xi^{L-2}$$
  
$$\vdots = \vdots$$

 $\sigma^L_+(\boldsymbol{R}(\xi)) = \boldsymbol{R}_L$ 

### Shift-and-cut and concatenability with zero

- $(\sigma_+(R)(\sigma)w)(0) = 0$
- $(\sigma_+^2(R)(\sigma)w)(0) = 0$ 
  - : = i
- $(\sigma_+^L(R)(\sigma)w)(0) = 0$

w is concatenable ⇔ with zero

 $col((\sigma_{+}^{i}(R))_{i=1,...,L}(\sigma))$  is a state map!

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#### Other equations equivalent to shift-and-cut ones $\implies$ different state maps are possible!

### Shift-and-cut and concatenability with zero

- $(\sigma_{+}(R)(\sigma)W)(0) = 0$ (-2(R)(-)W)(0) = 0
- $(\sigma_+^2(\boldsymbol{R})(\sigma)\boldsymbol{W})(\boldsymbol{0}) = \boldsymbol{0}$ 
  - : =:
- $(\sigma_+^L(\boldsymbol{R})(\sigma)\boldsymbol{w})(\mathbf{0}) = \mathbf{0}$

w is concatenable ⇔ with zero

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#### Other equations equivalent to shift-and-cut ones $\implies$ different state maps are possible!

Example: scalar systems

$$\mathbf{r}_0\mathbf{w} + \mathbf{r}_1\sigma\mathbf{w} + \ldots + \sigma^n\mathbf{w} = \mathbf{0}$$

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$$\mathbf{r}_0\mathbf{W} + \mathbf{r}_1\sigma\mathbf{W} + \ldots + \sigma^n\mathbf{W} = \mathbf{0}$$

Observe *w* concatenable with zero iff w = 0. Indeed,

$$\sigma_{+}^{n}(r)(\sigma)W = W$$
  

$$\sigma_{+}^{n-1}(r)(\sigma)W = r_{n-1}W + \sigma W$$
  

$$\vdots = \vdots$$
  

$$\sigma_{+}(r)(\sigma)W = r_{1}W + \ldots + \sigma^{n-1}W$$

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Zero at t = 0 iff  $(\sigma^k w)(0) = 0$  for k = 0, ..., n - 1.

### From kernel representation to state map

Denote 
$$\operatorname{col}((\sigma^i_+(R)))_{i=1,\dots,L} =: \Sigma_R$$
.

#### Theorem: Let $\mathfrak{B} = \ker(R(\sigma))$ . Then

$$R(\sigma)w = 0$$
  
$$\Sigma_R(\sigma)w = x$$

is a state representation of  $\mathfrak{B}$  with state variable x.

Algebraic characterization

**Theorem:** Let  $\mathfrak{B} = \ker(R(\sigma))$ , and define  $\Sigma_R$  as above. Then

$$\Xi_{R} := \{ f \in \mathbb{R}^{1 \times w}[\xi] \mid \exists g \in \mathbb{R}^{1 \times \bullet}[\xi], \alpha \in \mathbb{R}^{1 \times \bullet} \\ \text{s.t. } f = \alpha \Sigma_{R} + gR \}$$

is a vector space over  $\mathbb{R}$ .

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 $X \in \mathbb{R}^{\bullet \times w}[\xi]$  is state map for  $\mathfrak{B}$  iff row span $(X) = \Xi_R$ 

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is a vector space over  $\mathbb{R}$ .

 $X \in \mathbb{R}^{\bullet \times w}[\xi]$  is state map for  $\mathfrak{B}$  iff row span $(X) = \Xi_R$ 

X is minimal if and only if its rows are a basis for  $\Xi_R$ .

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

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  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\sigma_{+} \rightsquigarrow \begin{bmatrix} \boldsymbol{\xi} + \boldsymbol{2} & -\boldsymbol{1} \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + \boldsymbol{2} & -\boldsymbol{1} \end{bmatrix}$$

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\sigma_{+} \rightsquigarrow \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + 2 & -1 \end{bmatrix}$$
  
If  $(y, u) \in \mathfrak{B}$ , then for all  $g \in \mathbb{R}[\xi]$ 
$$[\sigma + 2 & -1] \begin{bmatrix} y \\ u \end{bmatrix} = [\sigma + 2 & -1] \begin{bmatrix} y \\ u \end{bmatrix}$$
$$+ \underbrace{g(\sigma) [\sigma^{2} + 2\sigma + 3 & -\sigma - 3]}_{=0 \text{ on } \mathfrak{B}} \begin{bmatrix} y \\ u \end{bmatrix}$$

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$$[\sigma + 2 & -1] \begin{bmatrix} y \\ u \end{bmatrix} = [\sigma + 2 & -1] \begin{bmatrix} y \\ u \end{bmatrix}$$
$$+ \underbrace{g(\sigma) [\sigma^{2} + 2\sigma + 3 & -\sigma - 3]}_{=0 \text{ on } \mathfrak{B}} \begin{bmatrix} y \\ u \end{bmatrix}$$

'equivalence modulo R'

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\begin{aligned} \sigma_+ &\leadsto \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + 2 & -1 \end{bmatrix} \\ \sigma_+^2 &\leadsto \begin{bmatrix} 1 & 0 \end{bmatrix} & \rightsquigarrow \begin{bmatrix} 1 & 0 \end{bmatrix} \end{aligned}$$

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\begin{aligned} \sigma_+ &\leadsto \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + 2 & -1 \end{bmatrix} \\ \sigma_+^2 &\leadsto \begin{bmatrix} 1 & 0 \end{bmatrix} & \rightsquigarrow \begin{bmatrix} 1 & 0 \end{bmatrix} \end{aligned}$$

If  $(y, u) \in \mathfrak{B}$ , then for all  $g \in \mathbb{R}[\xi]$ 

$$\begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} y \\ u \end{bmatrix} = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} y \\ u \end{bmatrix} + \underbrace{g(\sigma) [\sigma^2 + 2\sigma + 3 & -\sigma - 3]}_{=0 \text{ on } \mathfrak{B}} \begin{bmatrix} y \\ u \end{bmatrix}$$

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\begin{aligned} \sigma_+ &\leadsto \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + 2 & -1 \end{bmatrix} \\ \sigma_+^2 &\leadsto \begin{bmatrix} 1 & 0 \end{bmatrix} & \rightsquigarrow \begin{bmatrix} 1 & 0 \end{bmatrix} \end{aligned}$$

 $\Xi_{R} = \left\{ \alpha \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} + g(\xi) \begin{bmatrix} \xi^{2} + 2\xi + 3 & -\xi - 3 \end{bmatrix}, \\ \beta \begin{bmatrix} 1 & 0 \end{bmatrix} + f(\xi) \begin{bmatrix} \xi^{2} + 2\xi + 3 & -\xi - 3 \end{bmatrix} \\ \alpha, \beta \in \mathbb{R}, f, g \in \mathbb{R}[\xi] \right\}$ 

$$(\sigma^2 + 2\sigma + 3)y = (\sigma + 3)u$$
  $[\xi^2 + 2\xi + 3 - \xi - 3]$ 

$$\begin{aligned} \sigma_+ &\leadsto \begin{bmatrix} \xi + 2 & -1 \end{bmatrix} \rightsquigarrow \begin{bmatrix} \sigma + 2 & -1 \end{bmatrix} \\ \sigma_+^2 &\leadsto \begin{bmatrix} 1 & 0 \end{bmatrix} & \rightsquigarrow \begin{bmatrix} 1 & 0 \end{bmatrix} \end{aligned}$$

$$egin{aligned} \Xi_R &= \left\{ lpha \left[ eta+2 \quad -1 
ight] + eta(\xi) \left[ eta^2+2eta+3 \quad -eta-3 
ight], \ eta \left[ 1 \quad 0 
ight] + eta(\xi) \left[ eta^2+2eta+3 \quad -eta-3 
ight] \ lpha,eta \in \mathbb{R}, eta,eta \in \mathbb{R}, eta,eta \in \mathbb{R}[eta] 
ight\} \end{aligned}$$

Any set of generators of  $\Xi_R \rightsquigarrow$  a state map

## Outline

The axiom of state

Discrete-time systems

First-order representations

State maps

The shift-and-cut map

Algebraic characterization

#### Continuous-time systems

Computation of state-space representations

## On the space of solutions

$$\mathfrak{C}^{\infty}$$
-solutions to  $R(\frac{d}{dt})w = 0$  too small  $\rightsquigarrow \mathcal{L}_{1}^{\text{loc}}$ 

#### Equality in the sense of distributions:

$$R(\frac{d}{dt})w = 0 \qquad \Leftrightarrow \qquad \int_{-\infty}^{+\infty} w(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0$$
  
for all testing functions f.

- 1 - - -

### The axiom of state revisited

 $\Sigma = (\mathbb{T}, \mathbb{W}, \mathbb{X}, \mathfrak{B}_{full})$  is a state system if

$$(w_1, x_1), (w_2, x_2) \in \mathfrak{B}_{\text{full}} \text{ and } x_1(T) = x_2(T)$$
  
and  $x_1, x_2$  continuous at  $T$   
 $\downarrow$   
 $(w_1, x_1) \land (w_2, x_2) \in \mathfrak{B}_{\text{full}}$ 

'State map' 
$$\rightarrow X(\frac{d}{dt})$$

### From kernel representation to state map

Denote 
$$\operatorname{col}((\sigma^i_+(R)))_{i=1,\ldots,L} =: \frac{\Sigma_R}{\Sigma_R}$$
.

Theorem: Let  $\mathfrak{B} = \ker(R(\frac{d}{dt}))$ . Then

$$R(\frac{d}{dt})w = 0$$
  
$$\Sigma_R(\frac{d}{dt})w = x$$

is a state representation of  $\mathfrak{B}$  with state variable x.

#### ¿How to prove it?

$$\begin{array}{ll} 0 & \bigwedge_{0} w \in \mathfrak{B} & \Longleftrightarrow & \int_{-\infty}^{+\infty} (0 & \bigwedge_{0} w)(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \\ & \iff & \int_{0}^{+\infty} w(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \end{array}$$

for all testing functions f

$$\begin{array}{ll} 0 & \bigwedge_{0} w \in \mathfrak{B} & \Longleftrightarrow & \int_{-\infty}^{+\infty} (0 & \bigwedge_{0} w)(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \\ & \iff & \int_{0}^{+\infty} w(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \end{array}$$

#### for all testing functions f

Integrating repeatedly by parts on *f* yields:

$$\sum_{k=1}^{\deg(R)} \sum_{j=k}^{\deg(R)} (-1)^{k-1} (\frac{d^{j-k}}{dt^{j-k}} w)(0)^{\top} R_j^{\top} (\frac{d^{k-1}}{dt^{k-1}} f)(0) + \int_0^{+\infty} (R(\frac{d}{dt}) w)(t)^{\top} f(t) dt = 0$$
#### When is $\mathbf{w} \in \mathfrak{B}$ concatenable with zero?

$$\begin{array}{ll} 0 & \bigwedge_{0} w \in \mathfrak{B} & \Longleftrightarrow & \int_{-\infty}^{+\infty} (0 & \bigwedge_{0} w)(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \\ & \iff & \int_{0}^{+\infty} w(t)^{\top} (R(-\frac{d}{dt})^{\top} f)(t) dt = 0 \end{array}$$

#### for all testing functions f

Integrating repeatedly by parts on *f* yields:

$$\sum_{k=1}^{\deg(R)} \sum_{j=k}^{\deg(R)} (-1)^{k-1} (\frac{d^{j-k}}{dt^{j-k}} w)(0)^{\top} R_j^{\top} (\frac{d^{k-1}}{dt^{k-1}} f)(0) + \int_0^{+\infty} (R(\frac{d}{dt}) w)(t)^{\top} f(t) dt = 0$$

## $w \in \mathfrak{B}$ concatenable with zero if and only if...

$$\sum_{k=1}^{\deg(R)} \sum_{j=k}^{\deg(R)} (-1)^{k-1} (\frac{d^{j-k}}{dt^{j-k}} w) (0)^{\top} R_{j}^{\top} (\frac{d^{k-1}}{dt^{k-1}} f) (0) = 0$$

$$\begin{pmatrix} f(0) \\ (\frac{d}{dt} f) (0) \\ \vdots \\ (-1)^{\deg(R)-1} (\frac{d^{\deg(R)-1}}{dt^{\deg(R)-1}} f) (0) \end{bmatrix}^{\top} (\Sigma_{R} (\frac{d}{dt}) w) (0) = 0$$

$$\begin{pmatrix} (\Sigma_{R} (\frac{d}{dt}) w) (0) = 0 \end{pmatrix}$$

#### The shift-and-cut state map!

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From kernel representation to state representation

$$\pmb{R} \in \mathbb{R}^{g imes w}[\pmb{\xi}] \rightsquigarrow$$
 state map  $\pmb{X} \in \mathbb{R}^{n imes w}[\pmb{\xi}]$ 

Find:

$$\begin{split} \mathbf{E}, \mathbf{F} \in \mathbb{R}^{(n+g) \times n}, \, \mathbf{G} \in \mathbb{R}^{(n+g) \times w} \\ \mathbf{T} \in \mathbb{R}^{(n+g) \times g}[\boldsymbol{\xi}] \text{ with } \operatorname{rank}(\mathbf{T}(\lambda)) = g \, \forall \lambda \in \mathbb{C} \end{split}$$

satisfying

$$E\xi X(\xi) + FX(\xi) + G = T(\xi)R(\xi)$$

#### Linear equations, Gröbner bases computations!

#### From I/O representation to I/O/S representation

I/O representationstate map
$$R = \begin{bmatrix} P & -Q \end{bmatrix}$$
 $\sim$  $\begin{bmatrix} X_y & X_u \end{bmatrix}$ 

Find:

 $\boldsymbol{A} \in \mathbb{R}^{n \times n}, \, \boldsymbol{B} \in \mathbb{R}^{n \times m}, \, \boldsymbol{C} \in \mathbb{R}^{p \times p}, \, \boldsymbol{D} \in \mathbb{R}^{p \times m}$  $T \in \mathbb{R}^{(n+p) \times p}[\xi]$  with rank $(T(\lambda)) = g \ \forall \lambda \in \mathbb{C}$ 

satisfying

$$\begin{bmatrix} \xi X_y(\xi) & \xi X_u(\xi) \\ I_p & 0 \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} X_y(\xi) & X_u(\xi) \\ 0 & I_m \end{bmatrix} + T(\xi)R(\xi)$$

 $\sim$ 

State map + system equations

state-space equations

 $\sim$ 

State map + system equations

state-space equations

 $\sim$ 

State map + system equations

state-space equations

$$\left(\frac{d^2}{dt^2}+2\frac{d}{dt}+3\right)y=\left(\frac{d}{dt}+3\right)u$$

$$\begin{bmatrix} \xi^2 + 2\xi + 3 & -\xi - 3 \end{bmatrix}$$



$$A = \begin{bmatrix} -2 & 1 \\ -3 & 0 \end{bmatrix} \quad B = \begin{bmatrix} -1 \\ -3 \end{bmatrix}$$
$$C = \begin{bmatrix} 1 & 0 \end{bmatrix} \quad D = \begin{bmatrix} 0 \end{bmatrix}$$

'observer canonical form'



'observable canonical form'

• The state is constructed!

- The state is constructed!
- Axiom of state

- The state is constructed!
- Axiom of state
- Concatenability with zero

- The state is constructed!
- Axiom of state
- Concatenability with zero
- State maps

- The state is constructed!
- Axiom of state
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- State maps → state-space equations

- The state is constructed!
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- State maps → state-space equations
- Algorithms!