Bilinear differential forms and the Loewner framework for rational interpolation

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Linear theory *Leitmotiv*:

External structure mirrored in **internal structure**

Linearity & time-invariance → state & state equations

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External structure mirrored in **internal structure**

Linearity & time-invariance → state & state equations

Identification/Model order reduction:

- Ho-Kalman realization;
- Subspace identification;
- ...

IN THIS TALK:

External structure: bilinear on external variables

Internal structure: state equations

From data to state model

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HOW:

Loewner matrix → state trajectory → equations

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Loewner matrix ⇔ state trajectory **→** equations

factorization

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Loewner matrix state trajectory equations

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linear system solution

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External structure: bilinear on external variables

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From data to state model

HOW:

Loewner matrix → state trajectory → equations

Two-variable polynomial matrices

$$\frac{d}{dt}x = Ax + Bu \qquad \frac{d}{dt}x' = -A^{\top}x' + C^{\top}u'$$

$$y = Cx + Du \qquad y' = B^{\top}x' - D^{\top}u'$$

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$$u'^{\top}y + y'^{\top}u = \frac{d}{dt}(x'^{\top}x)$$

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$$u'^{\top}y + y'^{\top}u = \frac{d}{dt}(x'^{\top}x)$$

$$\begin{bmatrix} y' \\ u' \end{bmatrix} = \overline{w'}e^{\lambda}, \begin{bmatrix} u \\ y \end{bmatrix} = \overline{w}e^{\mu} \Longrightarrow x, x' \text{ vector-exponential}$$

$$\overline{\mathbf{w}'}^{\top}\overline{\mathbf{w}} = (\lambda + \mu)\overline{\mathbf{x}'}^{\top}\overline{\mathbf{x}}$$

$$\frac{d}{dt}x = Ax + Bu \qquad \frac{d}{dt}x' = -A^{\top}x' + C^{\top}u'$$

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$$u'^{\top}y + y'^{\top}u = \frac{d}{dt}(x'^{\top}x)$$

$$\begin{bmatrix} \mathbf{y}_i' \\ \mathbf{u}_i' \end{bmatrix} = \overline{\mathbf{w}_i'} \mathbf{e}^{\lambda_i \cdot}, \ \begin{bmatrix} \mathbf{u}_j \\ \mathbf{y}_i \end{bmatrix} = \overline{\mathbf{w}_j} \mathbf{e}^{\mu_j \cdot}, \ i, j = 1, \dots, N \Longrightarrow$$

$$\overline{\mathbf{w}_{i}^{\prime}}^{\top}\overline{\mathbf{w}_{i}}=(\lambda_{i}+\mu_{j})\overline{\mathbf{x}_{i}^{\prime}}^{\top}\overline{\mathbf{x}_{i}},\ i,j=1,\ldots,N$$

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$$\overline{\mathbf{w}_{i}'}^{\top} \overline{\mathbf{w}_{j}} = (\lambda_{i} + \mu_{j}) \overline{\mathbf{x}_{i}'}^{\top} \overline{\mathbf{x}_{j}}, \ \mathbf{i}, \mathbf{j} = 1, \dots, \mathbf{N}$$

$$\begin{bmatrix} \overline{\mathbf{w}_{i}'}^{\top} \overline{\mathbf{w}_{j}} \\ \lambda_{i} + \mu_{j} \end{bmatrix}_{\mathbf{i}, \mathbf{j} = 1, \dots, \mathbf{N}} = \begin{bmatrix} \overline{\mathbf{x}_{1}'} & \dots & \overline{\mathbf{x}_{N}'} \end{bmatrix}^{\top} \begin{bmatrix} \overline{\mathbf{x}_{1}} & \dots & \overline{\mathbf{x}_{N}} \end{bmatrix}$$
Loewner matrix \mathbb{L}

$$\frac{d}{dt}x = Ax + Bu \qquad \qquad \frac{d}{dt}x' = -A^{\top}x' + C^{\top}u'$$

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$$u'^{\top}y + y'^{\top}u = \frac{d}{dt}(x'^{\top}x)$$

$$\left[\frac{\overline{w_i'}^{\top}\overline{w_j}}{\lambda_i + \mu_j}\right]_{i,j = 1, \dots, N} = \left[\overline{x_1'} \quad \dots \quad \overline{x_N'}\right]^{\top} \left[\overline{x_1} \quad \dots \quad \overline{x_N}\right]$$

Loewner matrix L

Factorization ⇒ state trajectories

$$\frac{d}{dt}x = Ax + Bu \qquad \qquad \frac{d}{dt}x' = -A^{\top}x' + C^{\top}u'$$

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$$\begin{bmatrix} \overline{\mathbf{w}_{i}^{\prime\prime}}^{\top} \overline{\mathbf{w}_{i}} \\ \lambda_{i} + \mu_{i} \end{bmatrix}_{i,j=1,\dots,N} = \begin{bmatrix} \overline{\mathbf{x}_{1}^{\prime\prime}} & \dots & \overline{\mathbf{x}_{N}^{\prime\prime}} \end{bmatrix}^{\top} \begin{bmatrix} \overline{\mathbf{x}_{1}} & \dots & \overline{\mathbf{x}_{N}} \end{bmatrix}$$

For state equations solve for A, B, C, D

$$\begin{bmatrix} \mu_1 \overline{\mathbf{x_1}} & \dots & \mu_N \overline{\mathbf{x_N}} \\ \overline{\mathbf{y_1}} & \dots & \overline{\mathbf{y_N}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{C} & \mathbf{D} \end{bmatrix} \begin{bmatrix} \overline{\mathbf{x_1}} & \dots & \overline{\mathbf{x_N}} \\ \overline{\mathbf{u_1}} & \dots & \overline{\mathbf{u_N}} \end{bmatrix}$$

$$E\frac{d}{dt}x = Ax + Bu \qquad E^{\top}\frac{d}{dt}x' = -A^{\top}x' + C^{\top}u'$$

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$$\left[\frac{\overline{w_i'}^\top \overline{w_i}}{\lambda_i + \mu_i}\right]_{i,i=1,\ldots,N} = \begin{bmatrix} \overline{x_1'} & \ldots & \overline{x_N'} \end{bmatrix}^\top \boldsymbol{E} \begin{bmatrix} \overline{x_1} & \ldots & \overline{x_N} \end{bmatrix}$$

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$$u'^{\top} \left(\frac{d}{dt} \mathbf{y} \right) - \left(\frac{d}{dt} \mathbf{y}' \right)^{\top} \mathbf{u} = \frac{d}{dt} \left(\mathbf{x}'^{\top} \mathbf{A} \mathbf{x} \right)$$

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$$u'^{\top}\left(\frac{d}{dt}y\right) - \left(\frac{d}{dt}y'\right)^{\top}u = \frac{d}{dt}\left(x'^{\top}Ax\right)$$

$$\begin{bmatrix}
\overline{w_i'}^{\top} & \mathbf{0} & \mu_j \mathbf{I} \\
-\lambda_i \mathbf{I} & \mathbf{0}
\end{bmatrix}_{\overline{w_j}} \\
\lambda_i + \mu_j
\end{bmatrix}_{i,j=1,\dots,N} = \begin{bmatrix} \overline{x_1'} & \dots & \overline{x_N'} \end{bmatrix}^{\top} \mathbf{A} \begin{bmatrix} \overline{x_1} & \dots & \overline{x_N} \end{bmatrix}$$

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$$= \left[\overline{x_1'} \quad \dots \quad \overline{x_N'}\right]^{\top} A \left[\overline{x_1} \quad \dots \quad \overline{x_N}\right]$$
Shifted Loewner matrix \mathbb{L}_{σ}

For state equations, factor

$$\begin{bmatrix} \mathbb{L} & \mathbb{L}_{\sigma} \end{bmatrix} = X'^* \begin{bmatrix} \mathbf{E} \mathbf{X} & \mathbf{A} \mathbf{X} \end{bmatrix} \\ \begin{bmatrix} \mathbb{L} \\ \mathbb{L}_{\sigma} \end{bmatrix} = \begin{bmatrix} X'^* \mathbf{E} \\ \mathbf{X}'^* \mathbf{A} \end{bmatrix} \mathbf{X} .$$

For state equations, factor

$$\begin{bmatrix} \mathbb{L} & \mathbb{L}_{\sigma} \end{bmatrix} = X'^* \begin{bmatrix} EX & AX \end{bmatrix} \\ \begin{bmatrix} \mathbb{L} \\ \mathbb{L}_{\sigma} \end{bmatrix} = \begin{bmatrix} X'^*E \\ X'^*A \end{bmatrix} X.$$

If X, X'^{\top} have orthonormal rows (e.g. via SVD):

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If X, X'^{\top} have orthonormal rows (e.g. via SVD):

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Moreover,

$$B = -X' \begin{bmatrix} \overline{u_1'} & \dots & \overline{u_N'} \end{bmatrix}$$
 and $C = \begin{bmatrix} \overline{y_1} & \dots \overline{y_N} \end{bmatrix} X^{\top}$

Dual of
$$\mathfrak{B} = \operatorname{im} M\left(\frac{d}{dt}\right) = \ker R\left(\frac{d}{dt}\right)$$
:

$$\mathfrak{B}^{\perp} = \left\{ w' \in \mathfrak{C}^{\infty}(\mathbb{R}, \mathbb{R}^{\mathsf{w}}) \mid \int_{-\infty}^{+\infty} w'^{\top} w = 0 \right.$$
for all $w \in \mathfrak{B}$ of compact support

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x state for $w \in \mathfrak{B}$, x' for $w' \in \mathfrak{B}^{\perp}$:

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...same as 1st order case:

- Factorize Loewner matrix $\mathbb{L} = X'^{\top}X$;
- Solve for E, F, G

$$\begin{bmatrix} \mathbf{E} & \mathbf{F} & \mathbf{G} \end{bmatrix} \begin{bmatrix} \mathbf{X} \operatorname{diag}(\mu_i)_{i=1,\dots,N} \\ \mathbf{X} \\ \mathbf{W} \end{bmatrix} = \mathbf{0}$$

•
$$\mathfrak{B} = \{ w \mid \exists x \text{ s.t. } E \frac{d}{dt} x + Fx + Gw = 0 \}$$

Theorem: Assume N > n, the McMillan degree of the system.

If there are n linearly independent $w_i(\cdot) = \overline{w_i}e^{\mu_i \cdot}$, and $\mathbb{L} = X'^\top X$ is a rank-revealing factorization, then

 $\operatorname{rank} X = n$.

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Now $\sum_{i=1}^{n} \overline{x_i} \alpha_i = x(0)$, where $x(\cdot)$ is state trajectory of $w(\cdot) := \sum_{i=1}^{n} \overline{w_i} e^{\mu_i \cdot} \alpha_i$.

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 $w(\cdot)$ belongs to "autonomous" (w/out inputs) subbehavior $\Longrightarrow w(\cdot) = 0 \Longrightarrow$ contradiction.

Remarks

 As in subspace identification: compute state trajectories from data, solve for system matrices;

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 factorization ←→ state-space basis
- For J-self-dual systems, i.e. w.r.t. ⟨·,·⟩_J,
 L = X^TX is energy matrix ↔ storage function
 Lossless port-Hamilt./self-adj. Hamiltonian case
 (w/ Birthday Boy # 1)

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- For *J*-self-dual systems, i.e. w.r.t. $\langle \cdot, \cdot \rangle_J$, $\mathbb{L} = X^\top X$ is energy matrix \longleftrightarrow storage function Lossless port-Hamilt./self-adj. Hamiltonian case (w/ Birthday Boy # 1)
- Dual data computable from primal ones:

$$\overline{w}e^{\mu \cdot} \in \mathfrak{B} \text{ and } \overline{v}^{\top}\overline{w} = 0 \Longrightarrow \overline{v}e^{-\mu \cdot} \in \mathfrak{B}^{\perp}$$

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 General (non vector-exponential) discrete-time case (w/ Birthday Boy # 2) also possible.

Relations with interpolation

Left/right interpolation data:

$$\{(\mu_i, u_i'^*, y_i'^*) \in \mathbb{C} \times \mathbb{C}^p \times \mathbb{C}^m\}_{i=1,\dots,N}$$

$$\{(\lambda_i, u_i, y_i) \in \mathbb{C} \times \mathbb{C}^m \times \mathbb{C}^p\}_{i=1,\dots,N} .$$

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Looking for $H \in \mathbb{R}^{p \times m}(\xi)$ such that:

$$u_i^{\prime *} H(\mu_i) = y_i^{\prime *}, i = 1,..., N$$

 $H(\lambda_i) u_i = y_i, i = 1,..., N.$

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LCF/RCF $H(\xi) = N(\xi)D(\xi)^{-1} = P(\xi)^{-1}Q(\xi) \Longrightarrow$ model vector-exponential trajectories

$$\begin{bmatrix} u_i \\ y_i \end{bmatrix} e^{\mu_i \cdot}$$
 and $\begin{bmatrix} u'_i \\ y'_i \end{bmatrix} e^{\lambda_i \cdot}$

of

$$\begin{split} \mathfrak{B} &= \ker \; \left[\mathbf{\textit{Q}} \left(\frac{\textit{d}}{\textit{dt}} \right) \; \; - \mathbf{\textit{P}} \left(\frac{\textit{d}}{\textit{dt}} \right) \right] \; \text{and} \\ \mathfrak{B}^{\perp} &= \ker \; \left[\mathbf{\textit{D}}^{\top} \left(- \frac{\textit{d}}{\textit{dt}} \right) \; \; - \mathbf{\textit{N}}^{\top} \left(- \frac{\textit{d}}{\textit{dt}} \right) \right] \; , \end{split}$$

respectively.

$$\mathfrak{B} = \ker R\left(\frac{d}{dt}\right) = \operatorname{im} M\left(\frac{d}{dt}\right)$$

$$\mathfrak{B}^{\perp} = \operatorname{im} R^{\top}\left(-\frac{d}{dt}\right) = \ker M^{\top}\left(-\frac{d}{dt}\right)$$

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$$R(\xi)M(\xi)=0_{ exttt{p} imes exttt{m}}\Longrightarrow\exists\;\Psi(\zeta,\eta)\; ext{such that}$$
 $R(-\zeta)M(\eta)=(\zeta+\eta)\Psi(\zeta,\eta)$

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 $R(-\zeta)M(\eta) = (\zeta + \eta)\Psi(\zeta, \eta) = (\zeta + \eta)X'(\zeta)^{ op}X(\eta)$ with $X(\frac{d}{dt}), X'(\frac{d}{dt})$ state maps for \mathfrak{B} , resp. \mathfrak{B}^{\perp}

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$$\frac{R(-\zeta)M(\eta)}{\zeta+\eta} = X'(\zeta)^{\top}X(\eta) \iff \left[\frac{\overline{w_i'}^{\top}\overline{w_j}}{\lambda_i+\mu_j}\right]_{i,j=1,\dots,N} = X'^{\top}X$$

Data → state trajectories
 → state equations

using bilinear structure on state from bilinear structure on external variables

Data → state trajectories
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 using bilinear structure on state from bilinear structure on external variables

Model order reduction:
 Data → reduced-order state trajectories
 → reduced-order state equations

 Lossless port-Hamilt. & self-adj. Hamilt. case

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 Lossless port-Hamilt. & self-adj. Hamilt. case
- Current research:
 - 1D parametric modelling

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 using bilinear structure on state from bilinear structure on external variables
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 Data → reduced-order state trajectories
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 Lossless port-Hamilt. & self-adj. Hamilt. case
- Current research:
 - 1D parametric modelling
 - 2D case: Roesser models for $\overline{w_i}e^{\lambda_1^i \cdot 1}e^{\lambda_2^i \cdot 2}$, $i = 1, \dots, N$

Some related publications

- Rapisarda, P. and Trentelman, H.L. (2011) "Identification and data-driven model reduction of state-space representations of lossless and dissipative systems from noise-free data". *Automatica*, 47, (8), 1721-1728.
- van der Schaft, A.J. and Rapisarda, P. (2011) "State maps from integration by parts". SIAM J. Contr. Opt., 49, (6), 2415-2439.
- Rao, S. and Rapisarda, P. (2013) "Realization of lossless systems via constant matrix factorizations". *IEEE Trans. Aut. Contr.*, 58, (10), 2632-2636.
- Rapisarda, P. and van der Schaft, A.J. (2013) "Identification and data-driven reduced-order modeling for linear conservative port- and self-adjoint Hamiltonian systems". In *Proc. 52nd IEEE CDC*, Firenze, Italy.

THANK YOU

... and happy 60th birthdays, jongens!