The parametric Loewner Framework & the Curse-of-Dimensionality

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Problem Set-up

- Data: $s_i \in \mathbb{C}, \ \phi_i \in \mathbb{C}, \ i = 1, \dots, N.$
- **Find**: rational function $H(s) = \frac{n(s)}{d(s)}$,
 - s.t. $\mathbf{H}(s_i) = \phi_i, i = 1, \dots, N$.

• Omitted: Various generalizations of the Loewner framework to MIMO systems,

nonlinear systems, etc.

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Solution approaches:

- ▲ Barycentric framework (AAA)
- ▲ Loewner pencil

 $\rightarrow \left\{ \begin{array}{l} \text{Single-variable functions} \ \sim \ \text{linear systems} \\ \text{Multi-variate functions} \ \sim \ \begin{array}{l} \text{\textbf{parametrized}} \\ \text{linear systems} \end{array} \right.$

Lagrange basis, the Loewner matrix and rational interpolation

- Lagrange basis: Given $\lambda_i \in \mathbb{C}$, $\mathbf{q}_i(s) = \Pi_{i' \neq i}(s \lambda_{i'}), i = 1, \dots, n+1$.
- For given constants α_i , \mathbf{w}_i , $i=1,\cdots,n+1$, consider: $\sum_{i=1}^{n+1} \alpha_i \frac{\mathbf{g} \mathbf{w}_i}{s \lambda_i} = 0, \ \alpha_i \neq 0$
- Solving for **g** we obtain

$$\boxed{ \mathbf{g}(s) = \frac{\sum_{i=1}^{n+1} \frac{\alpha_i \mathbf{w}_i}{s - \lambda_i}}{\sum_{i=1}^{n+1} \frac{\alpha_i}{s - \lambda_i}} } \ = \ \frac{\sum_i \alpha_i \, \mathbf{w}_i \, \mathbf{q}_i(s)}{\sum_i \alpha_i \, \mathbf{q}_i(s)} \ \Rightarrow \ \mathbf{g}(\lambda_i) = \mathbf{w}_i, \ \forall \, \alpha_i,$$

This is the barycentric Lagrange interpolation formula.

The free parameters α_i , can be specified so that *additional* interpolation conditions are satisfied:

 $\mathbf{g}(\mu_j) = \mathbf{v}_j, \ j = 1, \dots, r$. For this to hold $\mathbb{L} \mathbf{c} = \mathbf{0}$, where

$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{v}_1 - \mathbf{w}_1}{\mu_1 - \lambda_1} & \cdots & \frac{\mathbf{v}_1 - \mathbf{w}_{n+1}}{\mu_1 - \lambda_{n+1}} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{v}_r - \mathbf{w}_1}{\mu_r - \lambda_1} & \cdots & \frac{\mathbf{v}_r - \mathbf{w}_{n+1}}{\mu_r - \lambda_{n+1}} \end{bmatrix} \in \mathbb{C}^{r \times (n+1)}, \quad \mathbf{c} = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_{n+1} \end{bmatrix} \in \mathbb{C}^{n+1}.$$

L: Loewner matrix with row array (μ_j, \mathbf{v}_j) , $j = 1, \dots, r$, and column array $(\lambda_i, \mathbf{w}_i)$, $i = 1, \dots, n+1$. The coefficients α_i are called barycentric weights.

Realization of the barycentric representation

Consider a rational function $\mathbf{H}(s)$, of the degree n. Let the Lagrange monomials be $\mathbf{x}_j = s - \lambda_j, \ \ j = 1, \cdots, n+1, \ \lambda_j \in \mathbb{C}$. Define:

• **pseudo-companion form** matrix of dimension $n \times (n+1)$:

$$\mathbb{X}(s) = \begin{bmatrix} x_1 & -x_2 & 0 & \cdots & 0 \\ x_1 & 0 & -x_3 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ x_1 & 0 & \cdots & 0 & -x_{n+1} \end{bmatrix} \in \mathbb{C}^{n \times (n+1)}[s],$$

- coefficient matrices: $\mathbb{A} = [\begin{array}{cccc} a_1 & a_2 & \cdots & a_{n+1} \end{array}], \quad \mathbb{B} = [\begin{array}{cccc} b_1 & b_2 & \cdots & b_{n+1} \end{array}] \in \mathbb{C}^{1 \times (n+1)}.$
- ullet \mathbb{A} : contains the barycentric weights and \mathbb{B} : the barycentric weights times the associated values of \mathbf{H} .

Theorem. Putting these quantities together we obtain a realization of H(s):

$$\Phi(s) = \underbrace{\begin{bmatrix} \mathbb{X}(s) & \mathbf{0} \\ \mathbb{A} & \mathbf{0} \\ \mathbb{B} & 1 \end{bmatrix}}_{(n+2)\times(n+2)}, \quad \mathbf{G} = \begin{bmatrix} \mathbf{0} \\ 1 \\ \mathbf{0} \end{bmatrix}, \quad \mathbf{W} = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{-1} \end{bmatrix} \Rightarrow \underbrace{\begin{bmatrix} \mathbf{H}(s) = \mathbf{W} \Phi(s)^{-1} \mathbf{G} \end{bmatrix}}_{\mathbf{H}(s) = \mathbf{W} \Phi(s)^{-1} \mathbf{G}}.$$

The realization $(\mathbf{w}, \mathbf{\Phi}, \mathbf{G})$ has dimension n+2, and is both R-controllable and R-observable, i.e. $[\mathbf{\Phi}, \mathbf{G}]$ and $\begin{bmatrix} \mathbf{H} \\ \mathbf{\Phi} \end{bmatrix}$, have full rank n+2, for all $s \in \mathbb{C}$.

The Loewner matrix - 2D interpolation

- $\mathcal{P}_{n,m}$: space of polynomials in two indeterminates, s and t, so that degree with respect to s is at most n and degree with respect to t is at most $m \Rightarrow \dim \mathcal{P}_{n,m} = (n+1)(m+1)$.
- Lagrange basis:

$$\mathbf{q}_{i,j}(s,t) = \prod_{i'\neq i} (s-\lambda_{i'}) \prod_{i'\neq i} (t-\pi_{j'}), \ i=1,\cdots,n+1, \ j=1,\cdots,m+1.$$

• Let $\mathbf{g}(s,t)$ satisfy: $\boxed{\sum_{i=1}^{n+1} \sum_{j=1}^{m+1} \ \alpha_{i,j} \ \frac{\mathbf{g} - \mathbf{w}_{i,j}}{(s - \lambda_i)(t - \pi_j)} = 0, \ \ \alpha_{i,j} \neq 0 } \quad \Rightarrow \quad$

$$\mathbf{g}(s,t) = \frac{\sum_{i=1}^{n+1} \sum_{j=1}^{m+1} \frac{\alpha_{i,j} \mathbf{w}_{i,j}}{(s-\lambda_i)(t-\pi_j)}}{\sum_{i=1}^{n+1} \sum_{j=1}^{m+1} \frac{\alpha_{i,j}}{(s-\lambda_i)(t-\pi_j)}} = \frac{\sum_{i=1}^{n+1} \sum_{j=1}^{m+1} \alpha_{i,j} \mathbf{w}_{i,j} \mathbf{q}_{i,j}(s,t)}{\sum_{i=1}^{n+1} \sum_{j=1}^{m+1} \alpha_{i,j} \mathbf{q}_{i,j}(s,t)}.$$

This is the **two-variable barycentric representation formula**. It follows that **g** satisfies the interpolation conditions $\mathbf{g}(\lambda_i, \pi_l) = \mathbf{w}_{i,l}$.

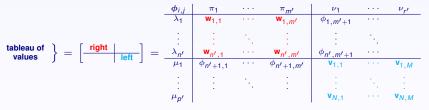
• The weights $\alpha_{i,j}$ can be determined so that **g** satisfies additional interpolation conditions:

$$\mathbf{g}(\mu_i, \nu_j) = \mathbf{v}_{i,j}, i = 1, \dots, p+1, j = 1, \dots, r+1,$$

where $(\mu_i, \nu_i; \mathbf{v}_{i,i})$ are given.

2D Loewner matrix

- Consider the arrays $\left\{ \begin{array}{l} P_c = \{(\lambda_j, \pi_i; \mathbf{w}_{j,i}): \ i=1,\cdots,n', \ j=1,\cdots,m' \\ P_r = \{(\mu_l, \nu_k; \mathbf{v}_{l,k}): \ k=1,\cdots,p', \ l=1,\cdots,r' \end{array} \right. .$
- The measurements can also be depicted as follows:



• The 2D Loewner matrix is:

$$\mathbb{L}_{\mathrm{2D}} = \begin{bmatrix} \frac{\mathbf{v}_{1,1} - \mathbf{w}_{1,1}}{(\mu_{1} - \lambda_{1})(\nu_{1} - \pi_{1})} & \frac{\mathbf{v}_{1,1} - \mathbf{w}_{1,2}}{(\mu_{1} - \lambda_{1})(\nu_{1} - \pi_{2})} & \cdots & \frac{\mathbf{v}_{1,1} - \mathbf{w}_{n',m'}}{(\mu_{1} - \lambda_{n'})(\nu_{1} - \pi_{m'})} \\ \frac{\mathbf{v}_{1,2} - \mathbf{w}_{1,1}}{(\mu_{1} - \lambda_{1})(\nu_{2} - \pi_{1})} & \frac{\mathbf{v}_{1,2} - \mathbf{w}_{1,2}}{(\mu_{1} - \lambda_{1})(\nu_{2} - \pi_{2})} & \cdots & \frac{\mathbf{v}_{1,2} - \mathbf{w}_{n',m'}}{(\mu_{1} - \lambda_{n'})(\nu_{2} - \pi_{m'})} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\mathbf{v}_{r',p'} - \mathbf{w}_{1,1}}{(\mu'_{r} - \lambda_{1})(\nu'_{p} - \pi_{1})} & \frac{\mathbf{v}_{r',p'} - \mathbf{w}_{1,2}}{(\mu'_{r} - \lambda_{1})(\nu'_{p} - \pi_{2})} & \cdots & \frac{\mathbf{v}_{r',p'} - \mathbf{w}_{n',m'}}{(\mu'_{r} - \lambda_{n'})(\nu'_{p} - \pi_{m'})} \end{bmatrix} \in \mathbb{C}^{p'r' \times n'm'}.$$

• Thus **g** satisfies the additional interpolation constraints given by P_r , if \mathbb{L}_{2D} **c** = 0.

Multivariate Loewner matrices and generalized Sylvester equations

The definition of \mathbb{L}_{ND} follows by means of Sylvester equations.

• N=2. Introduce the diagonal matrices $\Lambda, \Pi \in \mathbb{C}^{k_r q_r \times k_r q_r}$ and $M, N \in \mathbb{C}^{k_\ell q_\ell \times k_\ell q_\ell}$:

$$\Lambda = \operatorname{diag}(\lambda_1, \cdots, \lambda_{k_r}) \otimes \mathbf{I}_{q_r}, \ \Pi = \mathbf{I}_{k_r} \otimes \operatorname{diag}(\pi_1, \cdots, \pi_{q_r}),$$

$$\mathbf{M} = \operatorname{diag}(\mu_1, \cdots, \mu_{k_\ell}) \otimes \mathbf{I}_{q_\ell}, \ \mathbf{N} = \mathbf{I}_{k_\ell} \otimes \operatorname{diag}(\nu_1, \cdots, \nu_{q_\ell}).$$

Additionally, let $\mathbf{R} \in \mathbb{C}^{1 \times k_r q_r}$ and $\mathbf{L} \in \mathbb{C}^{k_\ell q_\ell \times 1}$ be vectors of ones.

Lemma. The 2D Loewner matrix \mathbb{L}_{2D} satisfies the **generalized Sylvester equation**:

$$\mathbf{N}\,\mathbf{M}\,\mathbb{L}_{2D} - \mathbf{N}\,\mathbb{L}_{2D}\,\boldsymbol{\Lambda} - \mathbf{M}\,\mathbb{L}_{2D}\,\boldsymbol{\Pi} + \mathbb{L}_{2D}\,\boldsymbol{\Lambda}\,\boldsymbol{\Pi} = \mathbb{V}\,\mathbf{R} - \mathbf{L}\,\mathbb{W} \tag{*}$$

Corollary. \mathbb{L}_{2D} , satsfies two standard coupled Sylvester equations:

$$NX - X\Pi = VR - LW$$
, $ML_{2D} - L_{2D}\Lambda = X$.

When X is eliminated the generalized Sylvester equation (*) results.

Loewner matrices L_{3D}

We introduce the **right data A**, Π , $\Theta \in \mathbb{C}^{k_r q_r v_r \times k_r q_r v_r}$, and the **left data M**, N, $Z \in \mathbb{C}^{k_\ell q_\ell v_\ell \times k_\ell q_\ell v_\ell}$:

$$\begin{split} & \boldsymbol{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_{k_r}) \otimes \boldsymbol{I}_{q_r} \otimes \boldsymbol{I}_{v_r}, & \boldsymbol{M} = \text{diag}(\mu_1, \dots, \mu_{k_\ell}) \otimes \boldsymbol{I}_{q_\ell} \otimes \boldsymbol{I}_{v_\ell}, \\ & \boldsymbol{\Pi} = \boldsymbol{I}_{k_r} \otimes \text{diag}(\pi_1, \dots, \pi_{q_r}) \otimes \boldsymbol{I}_{v_r}, & \boldsymbol{N} = \boldsymbol{I}_{k_\ell} \otimes \text{diag}(\nu_1, \dots, \nu_{q_\ell}) \otimes \boldsymbol{I}_{v_\ell}, \\ & \boldsymbol{\Theta} = \boldsymbol{I}_{k_r} \otimes \boldsymbol{I}_{q_r} \otimes \text{diag}(\theta_1, \dots, \theta_{v_r}), & \boldsymbol{Z} = \boldsymbol{I}_{k_\ell} \otimes \text{diag}(\zeta_1, \dots, \zeta_{v_\ell}). \end{split}$$

Let $\mathbf{R} \in \mathbb{C}^{1 \times k_r q_r v_r}$ and $\mathbf{L} \in \mathbb{C}^{k_\ell q_\ell v_\ell \times 1}$ be vectors of ones.

Lemma. The 3D Loewner matrix satisfies the **generalized Sylvester** equation:

$$\begin{split} Z \left(N\,M\,\mathbb{L}_{3D} - N\,\mathbb{L}_{3D}\,\Lambda - M\,\mathbb{L}_{3D}\,\Pi + \mathbb{L}_{3D}\,\Lambda\,\Pi \right) - \left(N\,M\,\mathbb{L}_{3D} - N\,\mathbb{L}_{3D}\,\Lambda - M\,\mathbb{L}_{3D}\,\Pi + \mathbb{L}_{3D}\,\Lambda\Pi \right) \Theta \\ &= \mathbb{V}\,R - L\,\mathbb{W}. \end{split}$$

Corollary. \mathbb{L}_{ND} satisfies a sequence of N coupled Sylvester equations:

1D:
$$\mathbb{L}_{1D} \Lambda - M \mathbb{L}_{1D} = \mathbb{V} \mathbf{R} - \mathbf{L} \mathbf{W},$$
2D:
$$\mathbf{N} \mathbb{X} - \mathbb{X} \Pi = \mathbb{V} \mathbf{R} - \mathbf{L} \mathbf{W}, \quad M \mathbb{L}_{2D} - \mathbb{L}_{2D} \Lambda = \mathbb{X},$$
3D:
$$\mathbf{Z} \mathbb{X}_1 - \mathbb{X}_1 \Theta = \mathbb{V} \mathbf{R} - \mathbf{L} \mathbf{W}, \quad \mathbf{N} \mathbb{X}_2 - \mathbb{X}_2 \Pi = \mathbb{X}_1, \quad \mathbf{M} \mathbb{L}_{3D} - \mathbb{L}_{3D} \Lambda = \mathbb{X}_2,$$

$$\vdots$$

Realization of mutlivariate rational functions

Consider a rational function **H** in N variables: ${}^i s, i = 1, \cdots, N$. Let the degree of **H** in each of these variables be $n_i, i = 1, \cdots, N$, and the Lagrange monomials in the variable ${}^i s$, be ${}^i \mathbf{x}_j = {}^i s - {}^i \lambda_j, \ j = 1, \cdots, n_i + 1, \ {}^i \lambda_j \in \mathbb{C}$. Associated with the i^{th} variable, we define the **pseudo-companion form** matrix:

$${}^{i}\mathbb{X} = \left[\begin{array}{ccccc} {}^{i}\mathbf{x}_{1} & {}^{-i}\mathbf{x}_{2} & 0 & \cdots & 0 \\ {}^{i}\mathbf{x}_{1} & 0 & {}^{-i}\mathbf{x}_{3} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ {}^{i}\mathbf{x}_{1} & 0 & \cdots & 0 & {}^{-i}\mathbf{x}_{n_{i}+1} \\ {}^{i}\epsilon_{1} & {}^{i}\epsilon_{2} & \cdots & {}^{i}\epsilon_{n_{i}} & {}^{i}\epsilon_{n_{i}+1} \end{array} \right] \in \mathbb{C}^{(n_{i}+1)\times(n_{i}+1)}[{}^{i}s].$$

The constants $^i\epsilon_j,\ j=1,\cdots,n_i+1,\$ are chosen so that $^i\mathbb{X}$ is **unimodular**.

Next, the variables are split into left (row) and right (column) variables. For simplicity, assume that ${}^{1}s, \cdots, {}^{k}s$ are the right variables and ${}^{k+1}s, \cdots, {}^{N}s$ are the left variables.

Define two Kronecker products: size $\kappa \times \kappa \left\{ \Gamma = {}^1\mathbb{X} \otimes \cdots \otimes {}^k\mathbb{X} \right\}$ and $\left\{ \Delta = {}^{k+1}\mathbb{X} \otimes \cdots \otimes {}^N\mathbb{X} \right\}$ size $\ell \times \ell$, where $\kappa = \prod_{i=1}^k (n_i + 1)$, and $\ell = \prod_{i=k+1}^N (n_i + 1)$. These matrices are unimodular.

Multi-row/multi-column indices and the coefficient matrices. Each column of Γ and each column of Δ defines a unique multi-index I_q , J_r . We will refer to these indices as **row-** and **column-multi-indices**:

$$I_q = [i_{k+1}^q, i_{k+2}^q, \cdots, i_N^q], J_r = [j_1^r, j_2^r, \cdots, j_k^r], q = 1, \cdots, \ell, r = 1, \cdots, \kappa.$$

Each multi-index $I_q(J_r)$ contains the indices of the Lagrange monomials involved in the $q^{th}(r^{th})$ column of $\Delta(\Gamma)$, respectively.

Remark. The ordering of these multi-indices is imposed by the ordering of the associated Kronecker products.

The coefficient matrices are:

$$\mathbb{A} = \left[\begin{array}{ccccc} a_{l_1,J_1} & a_{l_1,J_2} & \cdots & a_{l_1,J_{\kappa}} \\ a_{l_2,J_1} & a_{l_2,J_2} & \cdots & a_{l_2,J_{\kappa}} \\ \vdots & \vdots & \ddots & \vdots \\ a_{l_{\ell},J_1} & a_{l_{\ell},J_2} & \cdots & a_{l_{\ell},J_{\kappa}} \end{array} \right], \ \mathbb{B} = \left[\begin{array}{cccccc} b_{l_1,J_1} & b_{l_1,J_2} & \cdots & b_{l_1,J_{\kappa}} \\ b_{l_2,J_1} & b_{l_2,J_2} & \cdots & b_{l_2,J_{\kappa}} \\ \vdots & \vdots & \ddots & \vdots \\ b_{l_{\ell},J_1} & b_{l_{\ell},J_2} & \cdots & b_{l_{\ell},J_{\kappa}} \end{array} \right] \in \mathbb{C}^{\ell \times \kappa}.$$

A: contains the appropriately arranged barycentric weights of **H**.

B: contains the product of the barycentric weights times the associated values of H.

Theorem. Putting these quantities together we obtain a realization of $\mathbf{H}(^1s, \dots, ^Ns)$:

$$\boldsymbol{\Phi} = \begin{bmatrix} \boldsymbol{\Gamma}(1:\kappa-1,:) & \boldsymbol{0}_{\kappa-1,\ell-1} & \boldsymbol{0}_{\kappa-1,\ell} \\ & \boldsymbol{\mathbb{A}} & \boldsymbol{\Delta}(1:\ell-1,:)^T & \boldsymbol{0}_{\ell,\ell} \\ & \boldsymbol{\mathbb{B}} & \boldsymbol{0}_{\ell,\ell-1} & (\boldsymbol{\Delta})^T \end{bmatrix}, \; \boldsymbol{G} = \begin{bmatrix} \boldsymbol{0} \\ & \boldsymbol{\Delta}(\ell,:)^T \\ & & \boldsymbol{0} \end{bmatrix}, \; \boldsymbol{W} = [\boldsymbol{0}_{1,\kappa} \; \mid \; \boldsymbol{0}_{1,\ell-1} \; \mid \; -\boldsymbol{e}_{\ell}^T],$$

where \mathbf{e}_r denotes the r^{th} unit vector. Then: $\mathbf{H}(^1s, \dots, ^Ns) = \mathbf{W} \Phi(^1s, \dots, ^Ns)^{-1} \mathbf{G}$

$$\mathbf{H}(^1s,\cdots,^Ns)=\mathbf{W}\,\mathbf{\Phi}(^1s,\cdots,^Ns)^{-1}\,\mathbf{G}$$

 $(\mathbf{W}, \mathbf{\Phi}, \mathbf{G})$ has dimension $n = \kappa + 2\ell - 1$, and is both R-controllable and R-observable, i.e.

$$[\Phi,\; \textbf{G}] \;\; \text{and} \;\; \left[\begin{array}{c} \textbf{H} \\ \Phi \end{array} \right],$$

have full rank $\kappa + 2\ell - 1$, for all $s \in \mathbb{C}$. Furthermore Φ is a **polynomial matrix** in the variables is, while **W** and **G** are **constant**.

Remarks on the new realization of multivariate rational functions

• Since Δ is unimodular, using Schur complements, the dimension of the realization can be reduced to $\kappa+\ell-1$:

$$\hat{\boldsymbol{\Phi}} = \begin{bmatrix} \hline \Gamma(1:\kappa-1,:) & \boldsymbol{0}_{\kappa-1,\ell-1} \\ & \boldsymbol{\Delta} & \boldsymbol{\Delta}(1:\ell-1,:)^T \end{bmatrix}, \; \hat{\boldsymbol{G}} = \begin{bmatrix} \boldsymbol{0} \\ \hline \boldsymbol{\Delta}(\ell,:)^T \end{bmatrix}, \; \hat{\boldsymbol{W}} = \boldsymbol{e}_{\ell}^T \; \boldsymbol{\Delta}^{-T} \begin{bmatrix} \; \mathbb{B} \; \mid \; \boldsymbol{0}_{\ell,\ell-1} \; \end{bmatrix}.$$

This is achieved at the expense of introducing parameter dependence in \hat{W} .

• The above expression achieves the **multi-linearization** of the undelying **NEP** (nonlinear eigenvalue problem).

In the case of 2 variables and separation in left and right, we actually achieve a **linearization** of the underlying **NEP**.

- Key technical quantities: the **unimodular matrices** constructed for each variable.
- The possibility of splitting the variables to **left** variables and **right** variables, allows us to pick the splitting that **minimizes** n. For instance, if we have 4 variables with degrees 2, 2, 1, 1, splitting the variables into (2,1)–(2,1) gives n=17, while the splitting (2)–(2,1,1) (i.e. one column and 3 rows variables) gives n=26.
- It is conjectured that the size of the blue realization is the smallest possible with
 W and G constant (i.e. parameter independent).

Structure of the nullspace of ND-Loewner matrices

Issue: computing the **barycentric weights** $a_{i_1i_2...i_k}$, requires $\prod_{i=1}^N \nu_i^3$ flops.

Example with: n = 3, m = 2: $H(s, t) = \frac{s^3 t}{s - t^2 + 1}$, interpolation points chosen:

$$\left\{ \begin{array}{c|c|c|c} s_1 = 0 & s_2 = 1 & s_3 = 2 \\ t_1 = -\frac{1}{4} & t_2 = -\frac{3}{4} & t_3 = -\frac{5}{4} \\ \end{array} \right. \left. \begin{array}{c|c|c|c} s_4 = \frac{1}{2} & s_5 = \frac{3}{2} & s_6 = \frac{3}{4} \\ t_4 = -\frac{1}{2} & t_5 = -1 \\ \end{array} \right. \left. \begin{array}{c|c|c|c} s_6 = \frac{3}{4} & s_7 = \frac{2}{5} \\ t_7 = \frac{7}{2} & t_8 = \frac{7}{3} \\ \end{array} \right. = \left. \begin{array}{c|c|c|c} s_1 = 0 & s_2 = 1 \\ \end{array} \right.$$

$$\textbf{Tableau} = \begin{bmatrix} & t_1 & t_2 & t_3 & t_4 & t_5 & t_6 & t_7 & t_8 \\ \hline s_1 & h_{11} & h_{12} & h_{13} & h_{14} & h_{15} & \textbf{h}_{16} & h_{17} & h_{18} \\ s_2 & h_{21} & h_{22} & h_{23} & h_{24} & h_{25} & \textbf{h}_{26} & h_{27} & h_{28} \\ s_3 & h_{31} & h_{32} & h_{33} & h_{34} & h_{35} & \textbf{h}_{36} & h_{37} & h_{38} \\ s_4 & h_{41} & h_{42} & h_{43} & h_{44} & h_{45} & \textbf{h}_{46} & h_{47} & h_{48} \\ \hline s_5 & \textbf{h}_{51} & \textbf{h}_{52} & \textbf{h}_{53} & \textbf{h}_{54} & \textbf{h}_{55} & \textbf{h}_{56} & \textbf{h}_{57} & h_{58} \\ s_6 & \textbf{h}_{61} & \textbf{h}_{62} & \textbf{h}_{63} & \textbf{h}_{64} & \textbf{h}_{65} & \textbf{h}_{66} & \textbf{h}_{66} & h_{67} & h_{68} \\ s_7 & \textbf{h}_{71} & \textbf{h}_{72} & \textbf{h}_{73} & \textbf{h}_{74} & \textbf{h}_{75} & \textbf{h}_{76} & \textbf{h}_{77} & h_{78} \\ \hline s_8 & \textbf{h}_{81} & \textbf{h}_{82} & \textbf{h}_{83} & \textbf{h}_{84} & \textbf{h}_{85} & \textbf{h}_{86} & \textbf{h}_{86} & \textbf{h}_{86} & \textbf{h}_{87} & h_{88} \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -\frac{4}{31} & -\frac{12}{23} & -\frac{20}{7} & -\frac{2}{7} & -1 & 6 & -\frac{14}{41} & -\frac{21}{31} \\ -\frac{32}{47} & -\frac{32}{13} & -\frac{160}{23} & -\frac{16}{11} & -4 & -16 & -\frac{112}{37} & -\frac{84}{11} \\ -\frac{1}{4} & -\frac{1}{10} & \frac{5}{2} & -\frac{1}{20} & -\frac{1}{4} & \frac{1}{4} & -\frac{7}{7} & -\frac{21}{284} \\ -\frac{1}{6} & -\frac{1}{10} & \frac{5}{2} & -\frac{1}{20} & -\frac{1}{4} & \frac{1}{4} & -\frac{7}{72} & -\frac{281}{284} \\ -\frac{9}{26} & \frac{-81}{62} & -\frac{9}{2} & -\frac{9}{4} & -\frac{9}{4} & -\frac{81}{52} & -\frac{567}{212} \\ -\frac{1}{16} & -\frac{81}{304} & -\frac{16}{16} & -\frac{9}{64} & -\frac{9}{64} & -\frac{9}{64} & -\frac{9}{64} \\ -\frac{32}{2675} & -\frac{96}{1675} & \frac{32}{65} & -\frac{16}{575} & -\frac{4}{25} & \frac{425}{425} & -\frac{775}{775} & \frac{325}{325} \\ \frac{32}{1075} & \frac{32}{285} & -\frac{32}{385} & -\frac{16}{175} & -\frac{4}{25} & -\frac{16}{275} & -\frac{16}{5825} & -\frac{16}{2725} \end{bmatrix}$$

$$\begin{bmatrix} -2 & -2 & -\frac{54}{5} & -\frac{3}{4} & -1 & \frac{27}{20} & -\frac{32}{115} & -\frac{8}{15} & \frac{96}{425} & -\frac{32}{35} & \frac{8}{15} & \frac{32}{275} \\ 2 & -6 & -18 & \frac{3}{4} & -3 & \frac{9}{4} & \frac{32}{115} & -\frac{8}{5} & \frac{32}{35} & \frac{32}{35} & \frac{8}{5} & \frac{32}{165} \\ \frac{2}{3} & 6 & -54 & \frac{1}{4} & 3 & \frac{27}{4} & \frac{32}{345} & \frac{8}{5} & \frac{96}{95} & \frac{32}{105} & -\frac{8}{5} & \frac{32}{165} \\ -\frac{154}{31} & -\frac{526}{39} & -\frac{998}{31} & \frac{23}{124} & \frac{215}{93} & -\frac{2767}{10695} & \frac{32}{465} & -\frac{12752}{39525} & -\frac{4784}{7595} & \frac{32}{105} & -\frac{2416}{59565} \\ \frac{42}{23} & -\frac{318}{33} & -\frac{1210}{23} & \frac{561}{92} & \frac{15}{23} & -\frac{877}{97} & \frac{1136}{1136} & \frac{832}{345} & -\frac{8272}{345} & \frac{9872}{9872} & \frac{832}{635} & -\frac{2416}{5955} \\ -\frac{118}{21} & -\frac{34}{7} & -\frac{974}{7} & \frac{1217}{84} & \frac{277}{27} & -\frac{1847}{28} & \frac{15184}{2415} & \frac{1888}{105} & -\frac{35524}{245} & \frac{688}{245} & \frac{1886}{2655} \\ -\frac{26}{47} & \frac{590}{141} & \frac{7358}{235} & -\frac{325}{188} & -\frac{89}{705} & -\frac{1171}{171} & \frac{8824}{245} & -\frac{102}{235} & -\frac{7928}{9972} & \frac{6352}{635} & -\frac{68}{305} \\ \frac{1142}{13} & -\frac{13}{13} & \frac{39}{39} & \frac{260}{260} & -\frac{13}{13} & -\frac{780}{1495} & \frac{115}{1495} & -\frac{1154}{345} & \frac{1355}{245} & \frac{3908}{245} & -\frac{1245}{245} \\ \frac{1165}{235} & -\frac{2465}{245} & -\frac{2103}{235} & -\frac{12103}{169} & \frac{664}{495} & -\frac{1954}{245} & -\frac{3455}{245} & \frac{3152}{245} & -\frac{3908}{245} \\ \frac{13}{23} & -\frac{43}{23} & -\frac{169}{215} & -\frac{175}{29} & -\frac{199}{99} & \frac{379}{29} & \frac{28}{28} & \frac{121}{21} & -\frac{10532}{245} & -\frac{3164}{245} & \frac{1124}{215} & \frac{115}{265} & -\frac{115}{265} & -\frac{116}{265} & -\frac{115}{265} & -\frac{$$

 $\mathbb{L}_{3D} \in \mathbb{R}^{12 \times 12}$

The following observation holds:

$$\mathcal{L}_{S5} = \left[\begin{array}{cccc} \frac{21}{13} & \frac{33}{13} & \frac{207}{13} \\ \frac{69}{31} & \frac{117}{31} & \frac{783}{31} \\ 5 & 9 & 63 \end{array} \right], \ \mathcal{L}_{S6} = \left[\begin{array}{ccccc} \frac{5}{16} & \frac{2}{3} & -\frac{17}{16} \\ \frac{153}{304} & \frac{45}{38} & -\frac{621}{304} \\ \frac{57}{16} & 9 & -\frac{261}{16} \end{array} \right], \ \mathcal{L}_{S7} = \left[\begin{array}{ccccc} \frac{3904}{61525} & \frac{528}{2675} & -\frac{4544}{45475} \\ \frac{4544}{38525} & \frac{688}{1675} & -\frac{6484}{28475} \\ -\frac{5184}{7475} & -\frac{848}{3252} & \frac{8384}{5525} \end{array} \right],$$

$$L_{\mathbf{58}} = \left[\begin{array}{ccccc} -\frac{1856}{7525} & \frac{272}{1075} & \frac{832}{11825} \\ -\frac{832}{175} & \frac{144}{25} & \frac{1472}{825} \\ \frac{64}{275} & -\frac{592}{1925} & -\frac{192}{1925} \end{array} \right], \quad L_{l_6} = \left[\begin{array}{ccccccc} -\frac{27}{2} & \frac{27}{16} & \frac{24}{85} & \frac{8}{55} \\ -\frac{105}{2} & \frac{303}{16} & \frac{834}{85} & \frac{238}{55} \\ -\frac{2}{2} & \frac{16}{16} & \frac{85}{85} & -\frac{155}{165} \\ -\frac{41}{2} & \frac{65}{16} & \frac{233}{170} & \frac{113}{330} \end{array} \right] \Rightarrow$$

$$\mathcal{N}(\mathcal{L}_{S_6}) = \begin{bmatrix} 9 \\ -12 \\ 1 \end{bmatrix}$$

$$\mathcal{N}(\mathcal{L}_{S_6}) = \begin{bmatrix} -3 \\ 3 \\ 1 \end{bmatrix}$$

$$\mathcal{N}(\mathcal{L}_{S_6}) = \begin{bmatrix} -3 \\ 3 \\ 1 \end{bmatrix}$$

$$\mathcal{N}(\mathcal{L}_{S_7}) = \begin{bmatrix} -\frac{23}{17} \\ \frac{16}{17} \\ 1 \end{bmatrix}$$

$$\mathcal{N}(\mathcal{L}_{S_7}) = \begin{bmatrix} -\frac{7}{33} \\ -\frac{7}{33} \\ -\frac{16}{33} \\ 1 \end{bmatrix}$$

$$\mathcal{N}(\mathcal{L}_{S_8}) = \begin{bmatrix} -\frac{7}{33} \\ -\frac{16}{33} \\ 1 \end{bmatrix}$$

172 vs 1728 flops

⇒ decoupling

of the variables

Explanation: Constructing nullvectors of 2D from nullvectors of 1D Loewner matrices.

Consider the case n = 2, m = 2. The numerator and denominator can be expressed as:

$$\begin{array}{l} \text{num}^{2D} = \\ & a_{44}h_{44} \left(s - s_5\right)\left(s - s_6\right)\left(t - t_5\right)\left(t - t_6\right) + a_{45}h_{45} \left(s - s_5\right)\left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_5\right)\left(s - s_6\right)\left(t - t_4\right)\left(t - t_5\right) + a_{45}h_{54} \left(s - s_4\right)\left(s - s_6\right)\left(t - t_5\right)\left(t - t_6\right) + a_{45}h_{55} \left(s - s_4\right)\left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{56}h_{56} \left(s - s_4\right)\left(s - s_6\right)\left(t - t_4\right)\left(t - t_5\right) + a_{54}h_{64} \left(s - s_4\right)\left(s - s_5\right)\left(t - t_6\right)\left(t - t_6\right) + a_{55}h_{55} \left(s - s_4\right)\left(s - s_5\right)\left(t - t_4\right)\left(t - t_6\right) + a_{66}h_{66} \left(s - s_4\right)\left(s - s_5\right)\left(t - t_4\right)\left(t - t_5\right) + a_{44}h_{44} \left(s - s_4\right)\left(s - s_6\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{66} \left(s - s_4\right)\left(s - s_5\right)\left(t - t_4\right)\left(t - t_5\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_5\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_4\right)\left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_4\right)\left(t - t_6\right) + a_{46}h_{46} \left(s - s_6\right)\left(t - t_6\right) + a_{46}$$

The associated 1D rational functions are obtained by evaluating the 2D numerator/denominator at $s = s_i, t = t_i, i, j = 1, 2, 3$:

Example: $\mathbf{H}(s, t, x) = \frac{s^3 t^3 + x^2}{s^4 + x^4 t^2 + 2} \Rightarrow [\nu_1 \ \nu_2 \ \nu_3] = [5 \ 4 \ 3] \Rightarrow N = 60.$

Right points: $[s_1, s_2, s_3, s_4, s_5]$, $[t_1, t_2, t_3, t_4]$, $[x_1, x_2, x_3] \Rightarrow$

		5 4 3		
s ₁ , t ₁ , x ₁	s_2, t_1, x_1	s_3, t_1, x_1	s_4, t_1, x_1	s ₅ , t ₁ , x ₁
s_1, t_1, x_2	$\mathbf{s_2}, \mathbf{t_1}, \mathbf{x_2}$	s ₃ , t ₁ , x ₂	s_4, t_1, x_2	$\mathbf{s}_{5}, \mathbf{t}_{1}, \mathbf{x}_{2}$
$\mathbf{s_1},\mathbf{t_1},\mathbf{x_3}$	s_2, t_1, x_3	s_3, t_1, x_3	s_4, t_1, x_3	$\mathbf{s}_{5}, \mathbf{t}_{1}, \mathbf{x}_{3}$
s_1, t_2, x_1	s_2, t_2, x_1	s ₃ , t ₂ , x ₁	s_4, t_2, x_1	s ₅ , t ₂ , x ₁
${f s_1}, {f t_2}, {f x_2}$	s_2, t_2, x_2	s_3, t_2, x_2	s_4, t_2, x_2	$\mathbf{s}_{5},\mathbf{t}_{2},\mathbf{x}_{2}$
s_1, t_2, x_3	s_2, t_2, x_3	s_3, t_2, x_3	s_4, t_2, x_3	$\mathbf{s}_{5},\mathbf{t}_{2},\mathbf{x}_{3}$
s ₁ , t ₃ , x ₁	s ₂ , t ₃ , x ₁	s ₃ , t ₃ , x ₁	s ₄ , t ₃ , x ₁	s ₅ , t ₃ , x ₁
s_1, t_3, x_2	s_2, t_3, x_2	s_3, t_3, x_2	s_4, t_3, x_2	$\mathbf{s}_{5}, \mathbf{t}_{3}, \mathbf{x}_{2}$
s_1, t_3, x_3	s_2, t_3, x_3	s_3, t_3, x_3	s_4, t_3, x_3	s_5, t_3, x_3
s ₁ , t ₄ , x ₁	s_2, t_4, x_1	s_3, t_4, x_1	s_4, t_4, x_1	s_5, t_4, x_1
s_1, t_4, x_2	s_2, t_4, x_2	s_3, t_4, x_2	s_4, t_4, x_2	s_5, t_4, x_2
s_1, t_4, x_3	$\mathbf{s}_{2},\mathbf{t}_{4},\mathbf{x}_{3}$	s_3, t_4, x_3	s_4, t_4, x_3	$\mathbf{s}_5, \mathbf{t}_4, \mathbf{x}_3$

s_1, t_4, x_3	$\mathbf{s}_{2},\mathbf{t}_{4},\mathbf{x}_{3}$	s_3, t_4, x_3	s_4, t_4, x_3	$\mathbf{s}_{5},\mathbf{t}_{4},\mathbf{x}_{3}$
		5		
s_1, t_4, x_3	$\mathbf{s}_{2}, \mathbf{t}_{4}, \mathbf{x}_{3}$	s_3, t_4, x_3	$\mathbf{s_4}, \mathbf{t_4}, \mathbf{x_3}$	s_5, t_4, x_3

s₃, t₃, x₃

S1, t1, X3

 s_1, t_2, x_3

s₁, t₃, x₃

 s_2, t_1, x_3

 s_2, t_2, x_3

s2, t3, x3

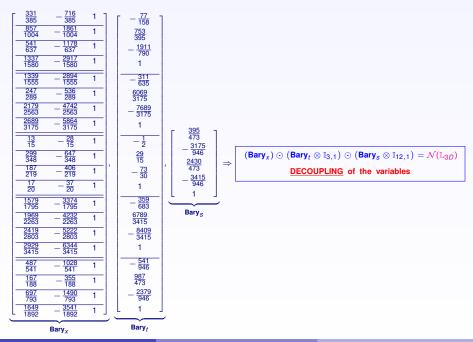
5 4 × 4 variable t

20 3 × 3 variable x

$a_{1,1,1}$	a _{2,1,1}	a _{3,1,1}	$a_{4,1,1}$	a _{5,1,1}
$a_{1,1,2}$	$a_{2,1,2}$	a _{3,1,2}	$a_{4,1,2}$	a _{5,1,2}
$a_{1,1,3}$	$a_{2,1,3}$	a _{3,1,3}	$a_{4,1,3}$	$a_{5,1,3}$
$a_{1,2,1}$	a _{2,2,1}	a _{3,2,1}	$a_{4,2,1}$	$a_{5,2,1}$
$a_{1,2,2}$	$a_{2,2,2}$	a _{3,2,2}	$a_{4,2,2}$	a _{5,2,2}
$a_{1,2,3}$	$\mathbf{a}_{2,2,3}$	$a_{3,2,3}$	$a_{4,2,3}$	$a_{5,2,3}$
$a_{1,3,1}$	$a_{2,3,1}$	$a_{3,3,1}$	$a_{4,3,1}$	a _{5,3,1}
$a_{1,3,2}$	a _{2,3,2}	a 3,3,2	$a_{4,3,2}$	a 5,3,2
$a_{1,3,3}$	$a_{2,3,3}$	$a_{3,3,3}$	$a_{4,3,3}$	$a_{5,3,3}$
$a_{1,4,1}$	$a_{2,4,1}$	a 3,4,1	$a_{4,4,1}$	a 5,4,1
$a_{1,4,2}$	$a_{2,4,2}$	a _{3,4,2}	$a_{4,4,2}$	a _{5,4,2}
$a_{1,4,3}$	a _{2,4,3}	a 3,4,3	$a_{4,4,3}$	a 5,4,3

Barv

s5, t3, x3



The ND-case. Consider $\mathbf{H}(\cdots, {}^{i}s, \cdots)$, of degrees $\nu_i - 1 > 0$, $i = 1, \cdots, N$. The following hold:

# of 1D Loewner matrices L	size of each $\mathbb L$	# of flops per L
$\nu_1 \nu_2 \cdots \nu_{N-2} \nu_{N-1}$	ν_N	ν_N^3
$\nu_1 \nu_2 \cdots \nu_{N-2}$	ν_{N-1}	ν_{N-1}^{3}
:	:	:
·	•	. 3
$\nu_1 \nu_2$	ν_3	$ u_3^3$
ν_1	ν_2	$ u_2^3$
1	ν_1	ν_1^3

Remark. The 1D Loewner matrices \mathbb{L}_{1D} , can be computed in parallel (simultanously).

Hence the total flops/storage required to compute an element of the null space of L_{ND} is:

$$\begin{aligned} & \textbf{Flops}_{1D} &= \nu_1^3 \, + \, (\nu_1) \, \nu_2^3 \, + \, (\nu_1 \nu_2) \, \nu_3^3 \, + \, \cdots \, + \, (\nu_1 \nu_2 \cdots \nu_{N-1}) \, \nu_N^3 & \text{vs.} & \textbf{Flops}_{ND} &= \nu_1^3 \nu_2^3 \cdots \nu_N^3, \\ & \textbf{Storage}_{1D} &= \nu_1^2 \, + \, (\nu_1) \, \nu_2^2 \, + \, (\nu_1 \nu_2) \, \nu_3^2 \, + \, \cdots \, + \, (\nu_1 \nu_2 \cdots \nu_{N-1}) \, \nu_N^2 & \text{vs.} & \textbf{Storage}_{ND} &= (\nu_1 \nu_2 \cdots \nu_N)^2. \end{aligned}$$

Resulting advantages:

(a) massive computational savings, making the procedure feasible.

Instead of computing ND-Loewner matrices of size $N = \Pi \nu_i$, one computes 1D-Loewner matrices of size ν_i , $i = 1, \dots, k$.

(b) improved numerical accuracy, again making this procedure feasible in many real situations.



Taming the curse of dimensionality

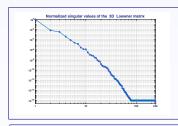
3D example: Numerical issues.

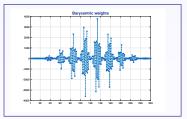
$$\mathbf{H}(s,t,x) = \frac{s^9 t^7 + s^3 + 5 x^2}{5 s^4 + 4 s^2 + x t^3 + 1} \ \Rightarrow \ \nu_1 = 10, \ \nu_2 = 8, \ \nu_3 = 3 \ \Rightarrow \ N = 240.$$

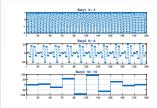
Flops for computing the barycentric weights: $10 \cdot 8 \cdot 3^3 + 10 \cdot 8^3 + 10^3 = 8$, 280 vs. $240^3 = 13$, 824, 000.

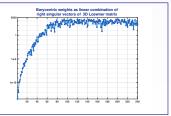
The nullspace of L_{3D} in floating point arithmetic, has dimension in excess of 100 instead of 1!

Compute the **barycentric weights** iteratively, by means of **91** 1D Loewner matrices.









Conclusions

- We extended the data-driven model reduction method, based of the Loewner Framework to N-parameter systems.
- The approach is closely related to the barycentric interpolation in N dimensions.
- Issue: complexity for many parameters, i.e. Curse-of-Dimensionality.
- Solution: reduce the procedure to one involving 1D Loewner matrices, only.

This addresses the Curse-of-Dimensionality (**C**-of-**D**) of multi-parameter linear systems by **decoupling** the variables. The computational complexity is thereby reduced by several orders of magnitude.

- In addition to computational effort in many cases this new method improves the numerical accuracy of the procedure.
- The data used are **ND-tensors**. Therefore we have provided a way to tame **C**-of-**D** for **ND-tensors**.

Reference:

• A.C. Antoulas, I.-V. Goşea, C. Poussot-Vassal, The Loewner framework for parametric systems: Taming the Curse-of-Dimensionality, https://arxiv.org/abs/2405.00495, May 2024. ... and in parting, a 20-D example, courtesy of Charles Poussot-Vassal:

$$\mathbf{H}(^{1}s,^{2}s,\cdots,^{20}s) =$$

$$\frac{3 \cdot {}^{1}s^{3} + 4 \cdot {}^{8}s + {}^{12}s + {}^{13}s \cdot {}^{14}s + {}^{15}s - {}^{20}s}{{}^{1}s + {}^{2}s^{2} \cdot {}^{3}s + {}^{4}s + {}^{5}s + {}^{6}s + {}^{7}s \cdot {}^{8}s + {}^{9}s \cdot {}^{10}s \cdot {}^{11}s + {}^{13}s + {}^{13}s^{3} + {}^{17}s + {}^{18}s \cdot {}^{19}s}$$

Statistics

- Complexity: (3,2,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1,1)
- 20-D tensor
- L_{20D} has size 6, 291, 456.
- 6,291,456² $\cdot \frac{8}{230} = 294,912$ GB of storage in double precision
- Full SVD: 2.49 · 10²⁰ flop

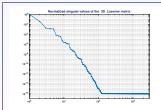
Cascaded SVD: $5.43 \cdot 10^7$ flop

Thanks for your attention

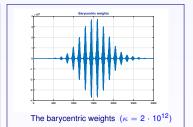
Appendix. A higher-degree example

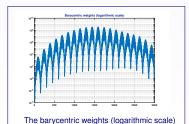
$$\mathbf{H}(s,t,x) = \frac{s^{19} t^{12} + s^3 + 5x}{5 s^{19} + 4 t^{14} + t^3 y^{9} + 1} \rightarrow \nu_1 = 20, \ \nu_2 = 15, \ \nu_3 = 10, \ N = 3000.$$

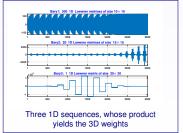
Number of flops required: $300 \cdot 10^3 + 20 \cdot 15^3 + 20^3 = 375,500$ vs $27 \cdot 10^9$ (ratio 1:71,904).



The normalized singular values of \mathbb{L}_{3D}







Appendix. Data usage: variables s, t, z, degrees $v_1 = 3, v_2 = 2, v_3 = 2$

		Z	1		z ₂			2		<i>z</i> ₃				z ₄					
	t ₁	t_2	t ₃	t ₄		t ₁	t ₂	t ₃	t ₄		t ₁	t ₂	t ₃	t_4		t ₁	t_2	t ₃	t ₄
s ₁	Х	X			s ₁	X	Х			s ₁					s ₁				
s ₂	X	X			s ₂	X	X			s ₂					s ₂				
s_3	X	X			<i>s</i> ₃	X	X			<i>s</i> ₃					s ₃				
<i>s</i> ₄					s ₄					<i>s</i> ₄			Х	X	S ₄			Х	Х
<i>s</i> ₅					<i>s</i> ₅					<i>s</i> ₅			x	X	<i>S</i> 5			X	X
<i>s</i> ₆]	<i>s</i> ₆					<i>s</i> ₆			X	X	<i>s</i> ₆			X	X

$$\left[\begin{array}{c|cccc} h_{112} & h_{122} & | & h_{212} & h_{222} & | & h_{312} & h_{322} \\ \underline{h_{132}} & h_{142} & | & \underline{h_{232}} & h_{242} & | & \underline{h_{332}} & h_{342} \\ \underline{\mathbb{L}_{1, \bullet, 2}} & & \underline{\mathbb{L}_{2, \bullet, 2}} & & \underline{\mathbb{L}_{3, \bullet, 2}} \end{array}\right] \left\{\begin{array}{c} \text{variable t}: & \textbf{3} & \mathbb{L}_{1D} \\ \textbf{12} \text{ evaluations (6 new)} \end{array}\right.$$

		Z	'n		z ₂				<i>z</i> ₃					<i>z</i> ₄					
	t ₁	t ₂	t ₃	t ₄		t ₁	t ₂	t ₃	t ₄		t ₁	t ₂	t ₃	t_4		t ₁	t ₂	t ₃	t ₄
<i>s</i> ₁					s ₁	Х	Х	Х	Х	s ₁					s ₁				
s ₂					s ₂	x	X	X	X	s ₂					s ₂				
<i>s</i> ₃					<i>s</i> ₃	x	X	x	X	<i>s</i> ₃					<i>s</i> ₃				
s ₄					s ₄					s ₄					s ₄				
<i>s</i> ₅					<i>s</i> ₅					<i>s</i> ₅					<i>s</i> ₅				
<i>s</i> ₆					<i>s</i> ₆					<i>s</i> ₆					<i>s</i> ₆				

```
\left[\begin{array}{ccc} h_{122} & h_{222} & h_{322} \\ \underline{h_{422}} & h_{522} & h_{622} \\ \underline{\mathbb{L}_{\bullet,2,2}} \end{array}\right] \left\{\begin{array}{ccc} \text{variable s}: & \textbf{1} & \mathbb{L}_{1D} \\ \text{6 evaluations (3 new)} \end{array}\right.
```

	<i>z</i> ₁					<i>z</i> ₂					<i>z</i> ₃						<i>z</i> ₄		
	t ₁	t_2	t ₃	t ₄		t ₁	t ₂	t ₃	t ₄		t_1	t_2	t ₃	t_4		t ₁	t_2	t ₃	t_4
s ₁					<i>s</i> ₁		X			s ₁					s ₁				
s ₂					s ₂		X			s ₂					s ₂				
s_3					<i>s</i> ₃		X			<i>s</i> ₃					<i>s</i> ₃				
S ₄					S ₄		X			S ₄					S ₄				
<i>s</i> ₅					<i>s</i> ₅		X			s ₅					<i>s</i> ₅				
<i>s</i> ₆					<i>s</i> ₆		X			<i>s</i> ₆					<i>s</i> ₆				

a		2		2		2		9044		2004		
a ₁₁₁ a ₁₁₂	1	a ₁₂₁ a ₁₂₂	1	a ₂₁₁ a ₂₁₂	1	a ₂₂₁ a ₂₂₂	1	a ₃₁₁ a ₃₁₂	1	a ₃₂₁ a ₃₂₂	1	Baryz
a ₁₁₂ a ₁₂₂	a ₁₁₂ a ₁₂₂	1	1	a ₂₁₂ a ₂₂₂	a ₂₁₂ a ₂₂₂	1	1	a ₃₁₂ a ₃₂₂	a ₃₁₂ a ₃₂₂	1	1	Bary _t
a ₁₂₂	a ₁₂₂	a ₁₂₂	a ₁₂₂	a ₂₂₂	a ₂₂₂	a ₂₂₂	a ₂₂₂	a ₃₂₂	a ₃₂₂	a ₃₂₂	a ₃₂₂	Barys
a ₁₁₁	a ₁₁₂	a ₁₂₁	a ₁₂₂	a ₂₁₁	a ₂₁₂	a ₂₂₁	a ₂₂₂	a ₃₁₁	a ₃₁₂	a ₃₂₁	a ₃₂₂	Bary _{tot}

Total number of function evaluations: 33.

Flops: 99 vs. 1728.