

Realification and PCC-Krylov Subspaces for Projection-Based Model Reduction

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Unreduced Model

Input-Output system represented as a system of Differential Algebraic Equations (DAEs)

$$\begin{array}{ccc} u_1(t) & \longrightarrow & \\ u_2(t) & \longrightarrow & \\ \vdots & & \\ u_p(t) & \longrightarrow & \end{array} \quad \boxed{\begin{array}{c} Ex' = Ax + Bu \\ y = B^T x, \end{array}} \quad \begin{array}{ccc} & \longrightarrow & y_1(t) \\ & \longrightarrow & y_2(t) \\ & & \vdots \\ & \longrightarrow & y_p(t) \end{array}$$

where $A, E \in \mathbb{R}^{N \times N}$ are sparse (possibly singular), $B \in \mathbb{R}^{N \times p}$.

- $N \gg p$ typically large, e.g. $N = \mathcal{O}(10^9)$
- $x(t) \in \mathbb{R}^N$ represents internal state.

Reduced Order Model (ROM) via Projection

System of DAEs of the same form

$$\begin{array}{ccc} u_1(t) & \longrightarrow & \\ u_2(t) & \longrightarrow & \\ \vdots & & \\ u_p(t) & \longrightarrow & \end{array} \quad \boxed{\begin{array}{c} \tilde{E}x' = \tilde{A}x + \tilde{B}u \\ y = \tilde{B}^T x \end{array}} \quad \begin{array}{ccc} & \longrightarrow & y_1(t) \\ & \longrightarrow & y_2(t) \\ & & \vdots \\ & \longrightarrow & y_p(t) \end{array}$$

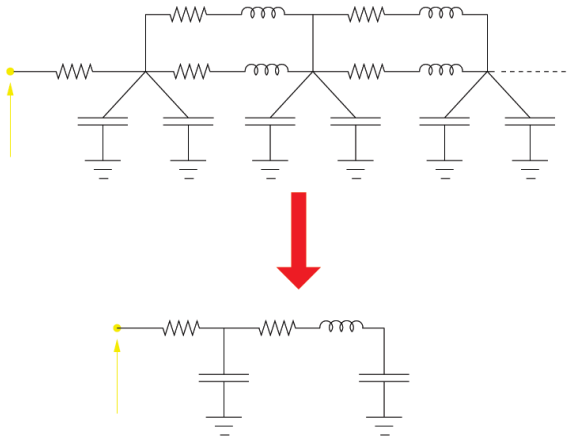
$$\tilde{A} := V^T A V, \quad \tilde{E} := V^T E V \quad \in \mathbb{R}^{n \times n}$$

$$\tilde{B} := V^T B \in \mathbb{R}^{n \times p},$$

with state-space dimension $n \ll N$ and $V \in \mathbb{R}^{N \times n}$ is basis for some ideal space.

Reduced Model Must Preserve Original Structure

Example: RCL circuit



Transfer Function

Relates Output directly to Input in Frequency Domain

Original system:

$$Ex' = Ax + Bu$$

$$y = B^T x.$$

Applying the Laplace transform,

$$sEX = AX(s) + BU(s)$$

$$Y(s) = B^T X(s).$$

In the frequency domain,

$$Y(s) = B^T (sE - A)^{-1} BU(s) \equiv H(s)U(s).$$

Transfer Function

Input \rightarrow Output Map in Frequency Domain

In the frequency domain, $Y(s) = H(s)U(s)$ with transfer function

$$H(s) = B^T (sE - A)^{-1} B \quad \in \quad (\mathbb{C} \cup \infty)^{p \times p}$$

For the reduced model,

$$\tilde{H}(s) = \tilde{B}^T (s\tilde{E} - \tilde{A})^{-1} \tilde{B} \quad \in \quad (\mathbb{C} \cup \infty)^{p \times p}$$

$$\tilde{H}(s) \approx H(s) \quad \Longleftrightarrow \quad \text{'Good' Reduced Order Model}$$

Moment Matching

Expressed as Taylor series expansion about $s_0 \in \mathbb{C}$:

$$\text{Original: } H(s) = \sum_{j=0}^{\infty} (s - s_0)^j M_j$$

$$\text{ROM: } \tilde{H}(s) = \sum_{j=0}^{\infty} (s - s_0)^j \tilde{M}_j$$

ROM matches n moments about σ if $\tilde{M}_j = M_j$ for $j = 1, 2, \dots, n$.

- Padè-Type Approximant

$$\tilde{H}(s) = H(s) + \mathcal{O}((s - s_0)^n)$$

Transfer function

Single-matrix formulation

Choose expansion point $s_0 \in \mathbb{C}$, re-write $H(s)$ as

$$\begin{aligned} H(s) &= B^T (sE - A)^{-1} B \\ &= B^T [(s - s_0)E + s_0E - A]^{-1} B \\ &= \textcolor{red}{B^T} (I - (s - s_0)H)^{-1} \textcolor{red}{R} \end{aligned}$$

(*Single matrix formulation*), where

$$H := -(s_0E - A)^{-1}E \quad \text{and} \quad R := (s_0E - A)^{-1}B.$$

We're interested in the *block-Krylov sequence*

$$R, HR, H^2R, \dots$$

Moments of the transfer function about s_0

Via Neumann (geometric series) expansion,

$$\begin{aligned} H(s) &= B^T (I - (s - s_0)H)^{-1} R \\ &= B^T \left(\sum_{j=0}^{\infty} (s - s_0)^j H^j \right) R \\ &= \sum_{j=0}^{\infty} (s - s_0)^j B^T H^j R \end{aligned}$$

- This the Taylor series expansion of $H(s)$ about s_0 .
- Block-Krylov sequence

$$R, HR, H^2R, \dots H^j R, \dots$$

Moment matching

...suggests n -th Block-Krylov subspace

$$\mathcal{K}_n(H, R) := \text{span} \{R, HR, H^2R, \dots, H^{n-1}R\}.$$

For $V \in \mathbb{R}^{N \times n}$ such that

$$\mathcal{K}_n(H, R) \subseteq \text{range } V,$$

ROM via projection on to V matches n moments about s_0 .

$$\tilde{B}^T \tilde{H}^j \tilde{R} = B^T H^j R \quad \text{for } j = 0, 1, 2, \dots, n-1$$

- because $\tilde{B}^T \tilde{H}^j \tilde{R} = B^T V \tilde{H}^j \tilde{R} = B^T H^j R$

Progressive Discovery of Krylov Space

For general $s_0 \in \mathbb{C}$,

$$H := -(s_0 E - A)^{-1} E \quad \text{and} \quad R := (s_0 E - A)^{-1} B$$

$$\mathcal{K}_n(H, R) := \text{span} \{R, HR, H^2 R, \dots, H^{n-1} R\} \subset \mathbb{C}^N$$

$H : \mathcal{K}_j(H, R) \rightarrow \mathcal{K}_{j+1}(H, R)$ *advances* the Krylov subspace.

- Not Real!
- We need $V \in \mathbb{R}^{N \times n}$ for which

$$\mathcal{K}_n(H, R) \subseteq \text{range } V.$$

Non-Real Expansion Point?

For $s_0 \in \mathbb{C}$, $H = H(s_0)$ and $R = R(s_0)$

Grimme (1997) suggested:

- Construct complex basis $V \in \mathbb{C}^{N \times \eta}$ for $\mathcal{K}_n(H, R)$
- Split V into (separate Re and Im parts)

$$V^* := \begin{bmatrix} v_1^r & v_1^i & v_2^r & v_2^i & \cdots & v_\eta^r & v_\eta^i \end{bmatrix} \in \mathbb{R}^{N \times 2\eta}$$

- Then,

$$\begin{aligned} \text{span } V^* &= \text{span } \mathcal{K}_n(H, R) \cup \mathcal{K}_n(\overline{H}, \overline{R}) \\ &= \text{span } \mathcal{K}_n(H(s_0), R(s_0)) \cup \mathcal{K}_n(H(\overline{s_0}), R(\overline{s_0})) \end{aligned}$$

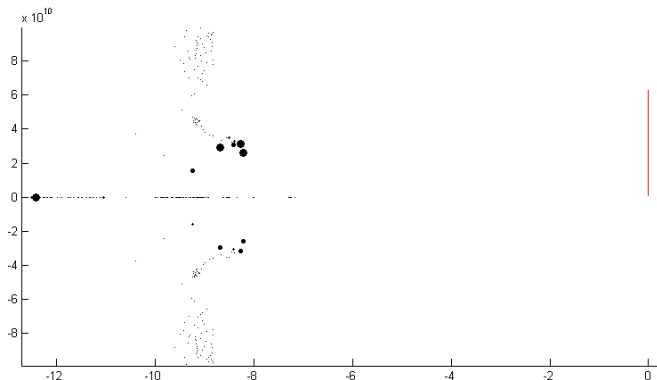
Match n moments about points s_0 and $\overline{s_0}$ simultaneously .

- $s_0 \notin \mathbb{R} \Rightarrow 2 \times$ model size, $4 \times$ computational cost
- Re-orthogonalize 2η vectors $v^* \in \mathbb{R}^N$? AIEEEE!

Motivation for Complex Expansion Point(s)

Example Pole Distribution

Poles of transfer function $H(s) = B^T(sE - A)^{-1}B$ correspond with eigenvalues of operator H .



We want eigen-information associated with dominant poles.

Complex Split

For a set of vectors $V \in \mathbb{C}^n$, the *split* of V is

$$V^* := V^{\mathbf{r}} \cup V^{\mathbf{i}} = \{ \operatorname{Re} v \cup \operatorname{Im} v \mid v \in V \} \subset \mathbb{R}^N.$$

Similarly for a vector space $S = \operatorname{span} V$,

$$\begin{aligned} S^* &:= \operatorname{span} V^* \\ &= \operatorname{span} V^{\mathbf{r}} \cup V^{\mathbf{i}} \\ &= \operatorname{span} V \cup \overline{V} \end{aligned}$$

(spans over \mathbb{C}).

Definition

Paired Complex Conjugate (PCC)-Krylov subspace induced by $H \in \mathbb{C}^{N \times N}$ and $R \in \mathbb{C}^N$.

$$\begin{aligned}\mathcal{K}_n(H, R)^* &= \text{span } \mathcal{K}_n(H, R) \cup \mathcal{K}_n(\overline{H}, \overline{R}) \\ &= \text{span } \left\{ R, \overline{R}, HR, \overline{HR}, H^2R, \overline{H^2R}, \dots, H^{n-1}R, \overline{H^{n-1}R} \right\}\end{aligned}$$

Equivalent Real Formulation

Definition

Realification functor on \mathbb{C}^N (Arnol'd, 1992).

$$\mathbb{R}\mathbb{C}^N = \mathbb{R}^{2N}$$

$$x + iy \in \mathbb{C}^N \quad \longleftrightarrow \quad \begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{R}^{2N}$$

For general $\{v_1, v_2, \dots, v_n\}$,

$$\mathbb{R} \left\{ \sum a_j v_j \mid a_j \in \mathbb{C} \right\} = \left\{ \sum a_j v_j \mid a_j \in \mathbb{R} \right\}.$$

- Similarly, *complexification* $\mathbb{C}(\mathbb{R}\mathbb{C}^N) = \mathbb{C}^N$

Equivalent Real Formulation

For operator matrix $H \in \mathbb{C}^{N \times N}$ and vector(s) $v \in \mathbb{C}^N$,

$$\hat{H} = \begin{bmatrix} H^{\mathbf{r}} & -H^{\mathbf{i}} \\ H^{\mathbf{i}} & H^{\mathbf{r}} \end{bmatrix} \in \mathbb{R}^{2N \times 2N} \quad \text{and} \quad \hat{v} = \begin{bmatrix} v^{\mathbf{r}} \\ v^{\mathbf{i}} \end{bmatrix} \in \mathbb{R}^{2N}.$$

$$w = \begin{bmatrix} w^{\mathbf{t}} \\ w^{\mathbf{b}} \end{bmatrix} \in \mathbb{R}\mathbb{C}^N \text{ has } \textit{top} \text{ and } \textit{bottom} \text{ parts in } \mathbb{R}^N.$$

Define the split of $W \subseteq \mathbb{R}\mathbb{C}^N$ as

$$W^* := W^{\mathbf{t}} \cup W^{\mathbf{b}} = \left\{ w^{\mathbf{t}} \cup w^{\mathbf{b}} \mid w \in W \right\} \subset \mathbb{R}^N.$$

Realified Krylov Subspace

Cheaper Basis Computation

Recall the Krylov subspace (over \mathbb{C})

$$\mathcal{K}_n(H, R) := \text{span} \{R, HR, H^2R, \dots, H^{n-1}R\} \subset \mathbb{C}^N.$$

Operator \hat{H} , initial vector block \hat{R} yield (over \mathbb{R})

$$\mathcal{K}_n(\hat{H}, \hat{R}) := \text{span} \{\hat{R}, \hat{H}\hat{R}, \hat{H}^2\hat{R}, \dots, \hat{H}^{n-1}\hat{R}\} \subset \mathbb{R}^{2N}.$$

Computing basis \hat{V} for $\mathcal{K}_n(\hat{H}, \hat{R})$ cuts cost in half.

- Why? Inner product in \mathbb{R}^{2N} vs. \mathbb{C}^N

Non-Equivalent Bases Yield The Same Split

Bases V and \widehat{V} of these isomorphic Krylov subspaces

$$\text{span } V = \mathcal{K}_n(H, R) \subset \mathbb{C}^N$$

$$\text{span } \widehat{V} = \mathcal{K}_n(\widehat{H}, \widehat{R}) \subset {}^{\mathbb{R}}\mathbb{C}^N$$

are not equivalent: ${}^{\mathbb{R}}V \neq \widehat{V}$ in general, but their splits

$$\text{span } V^* = \text{span } \widehat{V}^* = \mathcal{K}_n(H, R)^* \subset {}^{\mathbb{R}}\mathbb{C}^N$$

both span the same PCC-Krylov subspace.

- This follows from

$${}^{\mathbb{R}}(\cdot) : H^j R \leftrightarrow \widehat{H}^j \widehat{R}.$$

Concluding Remarks

- PCC-Krylov subspace is a complex space containing two complex conjugate Krylov subspaces and admits a real basis.
- Realified Krylov subspaces are cheaper to work with and ideal for methods implementing the general complex expansion point.