

The Extended-Rational Arnoldi Method

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1 Motivation

2 The Extended-Rational Krylov subspace

Linear Time-invariant system (LTI)

$$\begin{cases} \dot{x}(t) = A x(t) + B u(t) \\ y(t) = C x(t), \end{cases} \quad (1)$$

where $x(t) \in \mathbb{R}^n$ denotes the state vector and $u(t), y(t) \in \mathbb{R}^s$ are the input and output vectors respectively of the (LTI) system (1). The matrices $B, C^T \in \mathbb{R}^{n \times s}$ and $A \in \mathbb{R}^{n \times n}$.

In many applications, n has a high value $s \ll n$.

\implies It will be then reasonable to look for a reduced model having the form :

Reduced Model

$$\begin{cases} \dot{x}_m(t) &= A_m x_m(t) + B_m u(t) \\ y_m(t) &= C_m x_m(t), \end{cases} \quad (2)$$

with $A_m \in \mathbb{R}^{m \times m}$, $B_m, C_m^T \in \mathbb{R}^{m \times s}$, $x_m(t) \in \mathbb{R}^m$ and $y_m(t) \in \mathbb{R}^s$, with $m \ll n$ and such as :

- 1 $\|y - y_m\|$ is small.
- 2 The reduced model keeps the main properties of the original system (1).
- 3 The computational algorithm should be stable, and efficient.

Transfer Function

A classical way of relating the input to output by applying the Laplace transform

$$\mathcal{L}(f)(s) := \int_0^{\infty} e^{-st} f(t) dt,$$

in the system (1)

$$\begin{cases} sX(s) &= A X(s) + B U(s) \\ Y(s) &= C X(s) \end{cases},$$

where $X(s)$, $Y(s)$ and $U(s)$ are the Laplace transforms of $x(t)$, $y(t)$ and $u(t)$, respectively. Eliminating $X(s)$ in the previous two equations, we get

$$Y(s) = H(s) U(s),$$

where

$$H(s) = C (s I_n - A)^{-1} B. \quad (3)$$

The transfer function of the system (1).

- Approach the system (1) is the same as approach the function (3).

Among the most efficient techniques, there are the projection methods. Those methods are based on Krylov subspaces.

- The standard Krylov subspace :

$$\mathcal{K}_m(A, B) = \text{Vect}\{B, Ab, \dots, A^{m-1}B\}$$

- The Extended Krylov subspace :

$$\mathbb{K}_m(A, B) = \text{Range}\{A^{-m}B, \dots, A^{-1}B, B, AB, \dots, A^{m-1}B\}.$$

- The Rational Krylov subspace :

$$\mathbf{K}_m(A, B) = \text{Range}\{B, (A - s_2 I)^{-1}B, \dots, \prod_{i=2}^m (A - s_i I)^{-1}B\}$$

1 Motivation

2 The Extended-Rational Krylov subspace

New approach It consists on a new version more general than the Extended and Rational Krylov subspaces

for $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times s}$,

The extended rational Krylov subspace

$$\mathbb{K}_m^{er}(A, B) = \text{Range}\left\{\prod_{i=1}^m (A - s_i I_n)^{-1} B, \dots, (A - s_1 I_n)^{-1} B, B, \dots, A^{m-1} B\right\}.$$

where s_2, \dots, s_m are a chosen complex numbers

The Extended-Rational Arnoldi Algorithm

Algorithm 1 The Extended-Rational Arnoldi algorithm

- Inputs : $A \in \mathbb{R}^{n \times n}$, $V \in \mathbb{R}^{n \times s}$, $\{s_1, \dots, s_m\} \subset \mathbb{C}$ and m .
- Compute $[V_1, \Lambda] = QR([V, (A - s_1 I)^{-1} V])$, $\mathbb{V}_1 = [V_1]$.
- For $j = 1, \dots, m$
 - 1 Set $V_j^{(1)}$: first s columns of V_j ; $V_j^{(2)}$: second s columns of V_j .
 - 2 $\tilde{V}_{j+1} = [A V_j^{(1)}, (A - s_j I)^{-1} V_j^{(2)}]$.
 - 3 Orthogonalize \tilde{V}_{j+1} with respect to $\mathbb{V}_1, \dots, \mathbb{V}_j$ to get V_{j+1} , i.e.,
for $i = 1, 2, \dots, j$
$$H_{i,j} = (V_i)^T \tilde{V}_{j+1};$$
$$\tilde{V}_{j+1} = \tilde{V}_{j+1} - V_i H_{i,j};$$
end for
 - 4 $[V_{j+1}, H_{j+1,j}] = QR(\tilde{V}_{j+1})$.
 - 5 $\mathbb{V}_{j+1} = [\mathbb{V}_j, V_{j+1}]$.

End For.

The Extended-Rational Arnoldi Algorithm

$\mathbb{V}_m = [V_1, \dots, V_m] \in \mathbb{R}^{n \times 2ms}$ with columns constituting an orthonormal basis of the Extended-Rational block Krylov subspace $\mathbb{K}_m^{er}(A, B)$.

Upper block Hessenberg matrix $\overline{\mathbb{H}}_m$ whose non zero blocks are the $H_{i,j}$.

we also define :

- \mathbb{H}_m by removing the last blok of $\overline{\mathbb{H}}_m$
- $\mathbb{T}_m := \mathbb{V}_m^\top A \mathbb{V}_m$

The Extended-Rational Arnoldi Algorithm

The matrix \mathbb{T}_m is of great importance for the model reduction by the projection methods. When we manipulate big size model, the direct calculus of the matrix \mathbb{T}_m is of an elevated cost.

- lets $[B, (s_1 I - A)^{-1} B] = V_1 \Lambda$ be the QR decomposition of $[B, (s_1 I - A)^{-1} B]$

$$[B, (s_1 I - A)^{-1} B] = V_1 \Lambda = [V_1^{(1)}, V_1^{(2)}] \begin{bmatrix} \Lambda_{1,1} & \Lambda_{1,2} \\ 0 & \Lambda_{2,2} \end{bmatrix}.$$

- for $k = 1, \dots, m$, we assume that $H_{k+1,k}$ could be written as

$$H_{k+1,k} = \begin{bmatrix} H_{k+1,k}^{(1,1)} & H_{k+1,k}^{(1,2)} \\ 0 & H_{k+1,k}^{(2,2)} \end{bmatrix}.$$

Proposition

As defined previously $\overline{\mathbb{T}}_m$ and $\overline{\mathbb{H}}_m$ we could then :
for $k = 1, \dots, m$

$$\overline{\mathbb{T}}_m \tilde{\mathbf{e}}_{2k-1} = \begin{bmatrix} \overline{\mathbb{H}}_k \\ \mathbf{0}_{2(m-k)s \times 2ks} \end{bmatrix} = \overline{\mathbb{H}}_m \tilde{\mathbf{e}}_{2k-1},$$

$$\overline{\mathbb{T}}_m \tilde{\mathbf{e}}_2 = \left(\mathbf{s}_1 \begin{bmatrix} I_{2s} \\ \mathbf{0}_{2ms \times 2s} \end{bmatrix} \begin{bmatrix} \Lambda_{1,2} \\ \Lambda_{2,2} \end{bmatrix} - \overline{\mathbb{T}}_m \tilde{\mathbf{e}}_1 \Lambda_{1,2} - \tilde{\mathbf{e}}_1 \Lambda_{1,1} \right) (\Lambda_{2,2})^{-1},$$

and

$$\begin{aligned} \mathbb{T}_{m+1} \tilde{\mathbf{e}}_{2k+2} = & \left(\mathbf{s}_k \overline{\mathbb{H}}_m \tilde{\mathbf{e}}_{2k} - \begin{bmatrix} \overline{\mathbb{T}}_k \\ \mathbf{0}_{2(m-k)s \times 2ks} \end{bmatrix} \mathbb{H}_k \tilde{\mathbf{e}}_{2k} \right. \\ & \left. - \mathbb{T}_{m+1} \tilde{\mathbf{e}}_{2k+1} H_{k+1,k}^{(1,2)} - \tilde{\mathbf{e}}_{2k} \right) (H_{k+1,k}^{(2,2)})^{-1}, \end{aligned}$$

where $\tilde{\mathbf{e}}_i = \mathbf{e}_i \otimes I_{2s}$.

Useful relationships are satisfied by the Arnoldi procedure. Those equations have an important role for reduced model and for calculate the error norm.

Proposition

Soit $\bar{\mathbb{T}}_m = \mathbb{V}_{m+1}^T \mathbf{A} \mathbb{V}_m$ et $\mathbb{T}_m = \mathbb{V}_m^T \mathbf{A} \mathbb{V}_m$. Then we will have the following relation

$$\begin{aligned} \mathbf{A} \mathbb{V}_m &= \mathbb{V}_{m+1} \bar{\mathbb{T}}_m \\ &= \mathbb{V}_m \mathbb{T}_m + \mathbb{V}_{m+1} T_{m+1,m} \mathbb{E}_m^T, \end{aligned}$$

where $T_{m+1,m} = \mathbb{V}_{m+1}^T \mathbf{A} \mathbb{V}_m$ et $\mathbb{E}_m = [0_{2s \times 2(m-1)s}, I_{2s}]^T$.

Numerical Tests

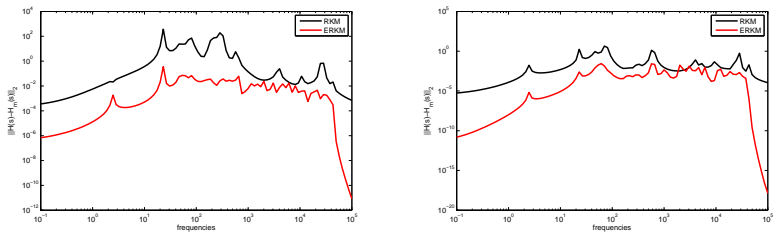


FIGURE – The CDp1ayer model. The norms of the errors $\|H(i\omega) - H_m(i\omega)\|_2$ for the Rational Block Arnoldi (black) and Extended-Rational Block Arnoldi (Red) for $\omega \in [10^{-1}, 10^5]$ and $s = 2$ with $m = 10$ (left) and $m = 20$ (right).

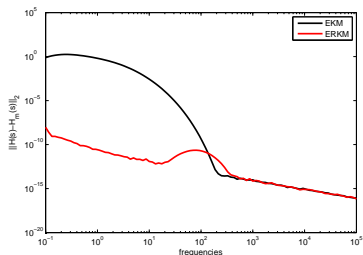
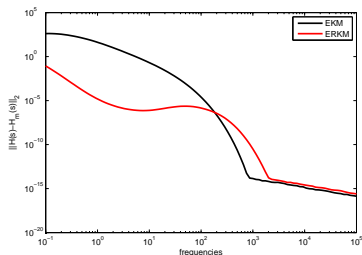


FIGURE – The Flow model ($n=9669$). The norms of the errors $\|H(i\omega) - H_m(i\omega)\|_2$ for the Extended Block Arnoldi (black) and Extended-Rational Block Arnoldi (Red) for the $\omega \in [10^{-1}, 10^5]$ and $s = 3$ with $m = 15$ (left) and $m = 30$ (right).

Thank you for your attention