Model reduction

Structure preserving model reduction of large-scale second-order linear dynamical systems

IT IAM

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. Manal Banda

Resulting

Higher

accuracy!

+ SOAR 24 + SOAR 24 + SOAR 24 + SOAR 22

97 98 99 10

97 98 99 1

- SOAB-\$4 - SOAB-\$4 - SOAB-12

ncy [MHz]

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Linearized thermoelasticity

Modeling of thermoelastic damping Second order linear dynamical systems Linearized thermoelasticity, which is the equations of motion coupled with the They appear in many physical systems of interest such as electrical, mechanical, electromechanically coupled, can be used to model an energy dissipation mechanism in MEMS called Thermoelastic damping. and thermomechanically coupled systems (MEMS). Energy loss from coupling of domains Single input single output system Thermal Domain Mechanical Domain $\mathbf{M}\ddot{\mathbf{z}}(t) + \mathbf{D}\dot{\mathbf{z}}(t) + \mathbf{K}\mathbf{z}(t) = \mathbf{b}u(t)$ Coupling Local Apply Time varying Time varying local stresses local strains Irreversible $y(t) = \mathbf{l}^* \mathbf{z}(t)$ external neratu in energy dissipation heat flow forces gradients SEM of 10.47 MHz lateral free-free beam resonator Transfer function Structure of equations $H(s) = \mathbf{l}^* \left(s^2 \mathbf{M} + s \mathbf{D} + \mathbf{K} \right)^{-1} \mathbf{b}$ Matrices involved are large. Computationally expensive to evaluate. M_{uu} $\begin{pmatrix} \ddot{\mathbf{u}} \\ \ddot{\boldsymbol{\theta}} \end{pmatrix} + \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{D}_{tu} & \mathbf{D}_{tt} \end{bmatrix} \begin{pmatrix} \dot{\mathbf{u}} \\ \dot{\boldsymbol{\theta}} \end{pmatrix} + \begin{bmatrix} \mathbf{K}_{uu} & \mathbf{K}_{ut} \\ \mathbf{0} & \mathbf{K}_{tt} \end{bmatrix} \begin{pmatrix} \mathbf{u} \\ \boldsymbol{\theta} \end{pmatrix}$ $\begin{pmatrix} \mathbf{b}_u \\ \mathbf{b}_t \end{pmatrix} u(t)$ = $\mathbf{M}_R = \mathbf{Y}^* \mathbf{M} \mathbf{X}$ $\mathbf{D}_R = \mathbf{Y}^* \mathbf{D} \mathbf{X}$ $\mathbf{b}_R = \mathbf{Y}^* \mathbf{b}$ pace X and Y which Select sm all projection subs accurately represents the motion in a certain $\mathbf{l}_R = \mathbf{X}^* \mathbf{l}$ $K_B = Y^*KX$ •Mass and damping matrices are singular and unsymmetric. ·Damping and stiffness matrices are unsymmetric $H_R(s) = \mathbf{l}_R^* \left(s^2 \mathbf{M}_R + s \mathbf{D}_R + \mathbf{K}_R \right)^{-1} \mathbf{b}_R$ Matrices involved are small!! •The coupling terms are tranposes of each other, $\mathbf{K}_{ut} = -\mathbf{D}_{tu}^T$ Computationally cheap to evaluate!! Proper selection of \mathbf{X} and \mathbf{Y} is crucial to preserve properties Structure preserving model reduction of the transfer function such as structure and approximation accuracy. Assume symmetric purely mechanical excitation Selection of projection subspace In practical applications, the devices of interest are excited purely mechanically in the form of electrostatic forces in the case of MEMS. Additionally, the displacement response at the loading point is usually the output quantity of interest. Second-order Krylov subspaces for X and Y Right and left second-order Krylov subspaces are related. Given matrices ${f A}$, ${f B}$ and starting vector ${f r}\,$, the kth second-order Krylov subspace $\mathcal{G}_k(\mathbf{A},\mathbf{B},\mathbf{r})$ is defined by the recursion relation, $\mathcal{G}_k\left(\mathbf{K}^{-1}\mathbf{D}, \mathbf{K}^{-1}\mathbf{M}; \mathbf{K}^{-1}\mathbf{b}\right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le i \le k-1} \right) \qquad \mathcal{G}_k\left(\mathbf{K}^{*,-1}\mathbf{D}^*, \mathbf{K}^{*,-1}\mathbf{M}^*; \mathbf{K}^{*,-1}\mathbf{l}\right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le i \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le i \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le i \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le i \le k-1} \right) = 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\end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right) = \operatorname{span}\left(\begin{bmatrix} \mathbf{r}_i^w \\ \mathbf{r}_i^t \end{bmatrix}_{0 \le k-1} \right)$ $\mathbf{r}_0 = \mathbf{r}$ By selecting \mathbf{X} and \mathbf{Y} as, k iterations of the Ar_0 \mathbf{r}_1 = $\mathbf{X} = \begin{bmatrix} \operatorname{span}\left\{\mathbf{r}_{0}^{u}, \dots, \mathbf{r}_{k-1}^{u}\right\} & \mathbf{0} \\ \mathbf{0} & \operatorname{span}\left\{\mathbf{r}_{0}^{t}, \dots, \mathbf{r}_{k-1}^{t}\right\} \end{bmatrix}$ SOAR iterations = $\mathbf{Ar}_{i-1} + \mathbf{Br}_{i-2} \ (2 \le i \le k-1)$ \mathbf{r}_i $Y = \overline{X}$ Select **x** to span the right second-order Krylov subspace **X** and **Y** contain the right and left second-order Krylov subspaces. $\mathcal{G}_k \left(\mathbf{K}^{-1} \mathbf{D}, \mathbf{K}^{-1} \mathbf{M}, \mathbf{K}^{-1} \mathbf{b} \right)$ •The structure of the equation is preserved. •By use of the theorem presented, one can match at least 2k moments by just Select Y to span the left second-order Krylov subspace producing a kth second-order Krylov subspace. This cannot be proven with the $\mathcal{G}_k\left(\mathbf{K}^{*,-1}\mathbf{D}^*,\mathbf{K}^{*,-1}\mathbf{M},\mathbf{K}^{*,-1}\mathbf{l}\right)$ version of the theorem presented for the first-order form of the transfer function. (Ref: R.-C. Li and Z. Bai, Structure-preserving model reduction using Krylov subspace formulation, 2005) All computational effort of the reduced order modeling is spent on generating these subspaces, i.e., conducting SOAR iterations. Numerical examples Moment matching theorem Michigan Free-Free beam structure. (Ref: W.-T. Hsu et al. Transducer 2001) Moments of the transfer function Material is Polysilicon Given a transfer function, the moments are defined as the coefficients of 2D plane stress analysis its power series expansion around a given point •Expand ROM at $s_0 = i9.5926 \times 10^6$ $H(s) = \sum_{i=1}^{n} M_i s^i$ Power series expansion around s=0. M_i ode at 100 [MHz] Schematic of structure and defor (Color represents temperature flu and M_{P_i} are the moments. Transfer function Dimensional equation Non-dimensional equation •Full model: 60318 DOFs Theorem * SOAR 24 * SOAR 24 * SOAR 22 •ROMs Let integers $k, r \ge 0$. If, ·SOAR 2: 2DOFs $\mathcal{G}_k\left(\mathbf{K}^{-1}\mathbf{D},\mathbf{K}^{-1}\mathbf{M};\mathbf{K}^{-1}\mathbf{b}\right)\subset \mathsf{span}\left(\mathbf{X}\right)\quad \mathsf{X} \text{ contains the kth right second-order Krylov subspace.}$ 2 SOAR iterations $\mathcal{G}_r\left(\mathbf{K}^{*,-1}\mathbf{D}^*,\mathbf{K}^{*,-1}\mathbf{M}^*;\mathbf{K}^{*,-1}\mathbf{l}\right) \subset \mathsf{span}\left(\mathbf{Y}\right) \quad \mathsf{Y} \text{ contains the rth left second-order Krylov subspace.}$ SOAR-S4: 4DOFs 2 SOAR iteration then Structure preserving model reduction The first k+r moments of the $M_i = M_{Bi}$ (0 < i < k + r - 1) transfer functions match 9.8 •SOAR 4: 4DOFs 4 SOAR iteration 9.5 95 and, Frequency [MHz] Frequ $H(s) = H_R(s) + O(s^{k+r})$ •SOAR-L2: 2DOFs 2 SOAR iterations for both left and right -\$838.4 -\$838.4 -\$938.4 Corollary second-order Krylov subspace for total of 4 The theorem above holds for a nonzero expansion point $s = s_0$ with minor SOAR iteration modification of the construction of the second-order Krylov subspaces SOAR-S4 is accurate irrespective of Plative **telative** moment matching property is a measure of the accuracy obtained from dimensionalization this Krylov subspace based reduced order modeling scheme. More matched ·SOAR-S4 requires less SOAR iterations. moments lead to a more accurate representation of the dynamical system. •SOAR-S4 requires 12 seconds to generate the transfer function (100 points) compared to 165 seconds for the full model. Frequency [MHz] Frequency [MHz]

I M E S