

AA216/CME345: PROJECTION-BASED MODEL ORDER REDUCTION

Balanced Truncation (BT)

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These slides are based on the recommended textbook: A.C. Antoulas, "Approximation of Large-Scale Dynamical Systems," Advances in Design and Control, SIAM, ISBN-0-89871-529-6

Outline

- 1 Reachability and Observability
- 2 Balancing
- 3 Balanced Truncation Method
- 4 Error Analysis
- 5 Stability Analysis
- 6 Computational Complexity
- 7 Comparison with the POD Method
- 8 Application
- 9 Balanced POD Method

- Consider the following **stable**, high-dimensional, LTI system

$$\begin{aligned}\frac{d\mathbf{w}}{dt}(t) &= \mathbf{A}\mathbf{w}(t) + \mathbf{B}\mathbf{u}(t) \\ \mathbf{y}(t) &= \mathbf{C}\mathbf{w}(t) \\ \mathbf{w}(0) &= \mathbf{w}_0\end{aligned}$$

- $\mathbf{w} \in \mathbb{R}^N$: State variables
 - $\mathbf{u} \in \mathbb{R}^{in}$: Input variables, typically $in \ll N$
 - $\mathbf{y} \in \mathbb{R}^q$: Output variables, typically $q \ll N$
- Recall that the solution $\mathbf{w}(t)$ of the above linear ODE can be written as

$$\mathbf{w}(t) = \phi(t, \mathbf{u}; t_0, \mathbf{w}_0) = e^{\mathbf{A}(t-t_0)}\mathbf{w}(t_0) + \int_{t_0}^t e^{\mathbf{A}(t-\tau)}\mathbf{B}\mathbf{u}(\tau)d\tau, \quad \forall t \geq t_0$$

(1)

Definition

A state $\mathbf{w} \in \mathbb{R}^N$ is **reachable** if there exist an input function $\mathbf{u}(\cdot)$ of finite energy and a time $T < \infty$ such that under this input and zero initial condition, the state of the system is \mathbf{w}

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A state $\mathbf{w} \in \mathbb{R}^N$ is **controllable** to the zero state if there exist an input function $\mathbf{u}(\cdot)$ and a time $T < \infty$ such that

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Definition

A state $\mathbf{w} \in \mathbb{R}^N$ is **unobservable** if for all $t \geq 0$,

$$\mathbf{y}(t) = \mathbf{C}\phi(t, \mathbf{0}; 0, \mathbf{w}) = \mathbf{0}_q$$

Definition (R.E. Kalman, 1963)

A linear dynamical system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ is **completely controllable** at time t_0 if it is not equivalent, for all $t \geq t_0$, to a system of the type

$$\frac{d\mathbf{w}^{(1)}}{dt} = \mathbf{A}^{(1,1)}\mathbf{w}^{(1)} + \mathbf{A}^{(1,2)}\mathbf{w}^{(2)} + \mathbf{B}^{(1)}\mathbf{u}$$

$$\frac{d\mathbf{w}^{(2)}}{dt} = \mathbf{A}^{(2,2)}\mathbf{w}^{(2)}$$

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- This definition can be extended to linear time-variant systems

Definition

A linear dynamical system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ is **completely observable** at time t_0 if it is not equivalent, for all $t \leq t_0$, to any system of the type

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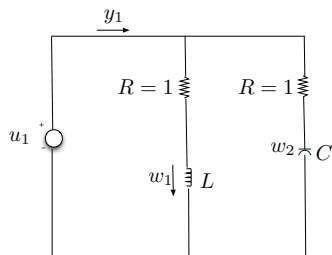
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- For $C = L$ and $R = 1$, the equation of the network shown above in terms of the current w_1 flowing through the inductor and the potential w_2 across the capacitor is given by

$$\frac{dw_1}{dt} = -\frac{1}{L}w_1 + u_1$$

$$\frac{dw_2}{dt} = -\frac{1}{L}w_2 + u_1$$

$$y_1 = \frac{1}{L}w_1 - \frac{1}{L}w_2 + u_1$$

- Under the change of variable $\tilde{w}_1 = (w_1 + w_2)/2$ and $\tilde{w}_2 = (w_1 - w_2)/2$, the previous dynamical system becomes

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- \tilde{w}_1 is controllable but not observable
- \tilde{w}_2 is observable but not controllable
- Hence, this dynamical system is neither completely controllable nor completely observable

Theorem (Kalman, 1961)

Consider a dynamical system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$. Then:

(i) There is a state space coordinate system in which the components of the state vector can be decomposed into four parts

$$\mathbf{w} = [\mathbf{w}^{(a)} \ \mathbf{w}^{(b)} \ \mathbf{w}^{(c)} \ \mathbf{w}^{(d)}]^T$$

(ii) The sizes N_a , N_b , N_c and N_d of these vectors do not depend on the choice of basis

(iii) The system matrices take the form

$$\mathbf{A} = \begin{bmatrix} \mathbf{A}^{(a,a)} & \mathbf{A}^{(a,b)} & \mathbf{A}^{(a,c)} & \mathbf{A}^{(a,d)} \\ 0 & \mathbf{A}^{(b,b)} & 0 & \mathbf{A}^{(b,d)} \\ 0 & 0 & \mathbf{A}^{(c,c)} & \mathbf{A}^{(c,d)} \\ 0 & 0 & 0 & \mathbf{A}^{(d,d)} \end{bmatrix}, \mathbf{B} = \begin{bmatrix} \mathbf{B}^{(a)} \\ \mathbf{B}^{(b)} \\ 0 \\ 0 \end{bmatrix},$$

$$\mathbf{C} = [0 \ \mathbf{C}^{(b)} \ 0 \ \mathbf{C}^{(d)}]$$

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- This theorem can be extended to linear time-variant systems

Definition

The **reachable subspace** $\mathbb{W}_{reach} \subset \mathbb{R}^N$ of a system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ is the set containing all reachable states of the system

$$\mathcal{R}(\mathbf{A} \ \mathbf{B}) = [\mathbf{B} \ \mathbf{A}\mathbf{B} \ \dots \ \mathbf{A}^{N-1}\mathbf{B}]$$

is the **reachability matrix** of the system

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Definition

The **controllable subspace** $\mathbb{W}_{contr} \subset \mathbb{R}^N$ of a system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ is the set containing all controllable states of the system

Theorem

Given a system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$,

$$\mathbb{W}_{reach} = \text{Im } \mathcal{R}(\mathbf{A}, \mathbf{B})$$

■ Proof

- recall (1), set $t_0 = 0$ and $\mathbf{w}(0) = \mathbf{0}$, and consider the impact of applying unit inputs at $t = 1$, $t = 2$, and so on
- recall that $e^{\mathbf{A}t} = \mathbf{I}_N + \mathbf{A}t + \frac{(\mathbf{A}t)^2}{2!} + \frac{(\mathbf{A}t)^3}{3!} + \dots$
- input at $t = 1$: $u(\tau) = \delta(\tau - 1) \Rightarrow \mathbf{B}, e^{\mathbf{A}}\mathbf{B}, e^{2\mathbf{A}}\mathbf{B}, \dots$
- input at $t = 2$: $u(\tau) = \delta(\tau - 2) \Rightarrow \mathbf{B}, e^{\mathbf{A}}\mathbf{B}, \dots$
- and so on
- this leads to the formulation $\mathcal{R} = [\mathbf{B} \ \mathbf{A}\mathbf{B} \ \dots \ \mathbf{A}^{N-1}\mathbf{B}]$
- each $\mathbf{A}\mathbf{B}$ term captures the state influence of applying control inputs at previous time horizons offset by dynamics defined by \mathbf{A}

Corollary

(i) If \mathcal{R} has full rank, $\mathbf{A}\mathbb{W}_{reach} \subset \mathbb{W}_{reach}$

(ii) The system is completely reachable if and only if $\text{rank } \mathcal{R}(\mathbf{A}, \mathbf{B}) = N$

(iii) Reachability is basis independent

■ Proof

- only the term $\mathbf{A}^N \mathbf{B} \in \mathbb{R}^{N \times in}$ requires special attention (Hint: use the fact that \mathcal{R} has full (row) rank)

Definition

The **reachability (controllability) Gramian** at time $t < \infty$ is defined as the $N \times N$ symmetric positive semi-definite matrix

$$\mathcal{P}(t) = \int_0^t e^{\mathbf{A}\tau} \mathbf{B} \mathbf{B}^* e^{\mathbf{A}^* \tau} d\tau$$

where \star designates the transpose of the complex conjugate

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Definition

The **observability Gramian** at time $t < \infty$ is defined as the $N \times N$ symmetric positive semi-definite matrix

$$\mathcal{Q}(t) = \int_0^t e^{\mathbf{A}^* \tau} \mathbf{C}^* \mathbf{C} e^{\mathbf{A}\tau} d\tau$$

Proposition

The columns of $\mathcal{P}(t)$ span the reachability subspace

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Corollary

A system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ is reachable if and only if $\mathcal{P}(t)$ is symmetric positive definite at some time $t > 0$

Theorem

For continuous linear dynamical systems, the notions of controllability and reachability are equivalent – that is,

$$\mathbb{W}_{reach} = \mathbb{W}_{contr}$$

Definition

The **unobservability subspace** $\mathbb{W}_{unobs} \subset \mathbb{R}^N$ is the set of all unobservable states of the system and the matrix

$$\mathcal{O}(\mathbf{C}, \mathbf{A}) = [\mathbf{C}^* \ \mathbf{A}^* \mathbf{C}^* \ \dots \ (\mathbf{A}^*)^j \mathbf{C}^* \ \dots]^*$$

is the **observability matrix** of the system

Theorem

Given a system $(\mathbf{A}, \mathbf{B}, \mathbf{C})$,

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- (i) $\mathbf{A}\mathbb{W}_{unobs} \subset \mathbb{W}_{unobs}$
- (ii) The system is completely observable if and only if $\text{rank } \mathcal{O}(\mathbf{C}, \mathbf{A}) = N$
- (iii) Observability is basis independent

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The **infinite reachability (controllability) Gramian** is defined for a **stable** LTI system as the $N \times N$ symmetric positive semi-definite matrix

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The **infinite observability Gramian** is defined for a **stable** LTI system as the $N \times N$ symmetric positive semi-definite matrix

$$\mathcal{Q} = \int_0^{\infty} e^{\mathbf{A}^* t} \mathbf{C}^* \mathbf{C} e^{\mathbf{A}t} dt$$

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$$\mathcal{P} = \frac{1}{2\pi} \int_{-\infty}^{\infty} (j\omega \mathbf{I}_N - \mathbf{A})^{-1} \mathbf{B} \mathbf{B}^* (-j\omega \mathbf{I}_N - \mathbf{A}^*)^{-1} d\omega$$

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$$\mathbf{A}^*\mathcal{Q} + \mathcal{Q}\mathbf{A} + \mathbf{C}^*\mathbf{C} = \mathbf{0}_N$$

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$$\|\mathbf{w}\|_{\mathcal{P}^{-1}}^2 = \mathbf{w}^T \mathcal{P}^{-1} \mathbf{w} \quad \left(\leq \int_0^t (\mathbf{B}\mathbf{u}(\tau))^* \mathbf{B}\mathbf{u}(\tau) d\tau \right)$$

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- The inner product based on \mathcal{Q} indicates the maximal energy produced by observing the output of the system corresponding to an initial state \mathbf{w}_0 when no input is applied

$$\|\mathbf{w}\|_{\mathcal{Q}}^2 = \mathbf{w}^T \mathcal{Q} \mathbf{w}$$

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└ Model Order Reduction Based on Balancing

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 - **difficult to observe**, i.e., produce a small observation energy $\|\mathbf{w}\|_{\mathcal{Q}}^2$
- The above notions are **basis-dependent**
- One would like to consider a basis where both concepts are **equivalent**, i.e., where the system is **balanced**

└ Balancing

└ The Effect of Basis Change on the Gramians

- Balancing requires changing the basis for the state using a transformation $\mathbf{T} \in \text{GL}(N)$

$$\tilde{\mathbf{w}} = \mathbf{T}\mathbf{w}$$

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- Then (see Chapter 5)

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- Then (see Chapter 5)
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└ The Effect of Basis Change on the Gramians

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- the observability Gramian becomes

$$\tilde{\mathcal{Q}} = \mathbf{T}^{*-1}\mathcal{Q}\mathbf{T}^{-1}$$

Balancing

Balancing Transformation

- The balancing transformations \mathbf{T}_{bal} and $\mathbf{T}_{\text{bal}}^{-1}$ can be computed as follows

1 compute the Cholesky factorization $\mathcal{P} = \mathbf{U}\mathbf{U}^*$

2 compute the eigenvalue decomposition of $\mathbf{U}^* \mathbf{Q} \mathbf{U}$

$$\mathbf{U}^* \mathbf{Q} \mathbf{U} = \mathbf{K} \mathbf{\Sigma}^2 \mathbf{K}^*$$

where the entries in $\mathbf{\Sigma}$ are ordered decreasingly

3 compute the transformations

$$\mathbf{T}_{\text{bal}} = \mathbf{\Sigma}^{\frac{1}{2}} \mathbf{K}^* \mathbf{U}^{-1}$$

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- Then, one can check that balancing is achieved

$$\mathbf{T}_{\text{bal}} \mathcal{P} \mathbf{T}_{\text{bal}}^* = \mathbf{T}_{\text{bal}}^{*-1} \mathcal{Q} \mathbf{T}_{\text{bal}}^{-1} = \mathbf{\Sigma}$$

Definition (Hankel Singular Values)

$\mathbf{\Sigma} = \text{diag}(\sigma_1, \dots, \sigma_N)$ contains the N **Hankel singular values** of the system

- Computing the balancing transformation \mathbf{T}_{bal} is equivalent to minimizing the following function

$$\min_{\mathbf{T} \in \text{GL}(N)} f(\mathbf{T}) = \min_{\mathbf{T} \in \text{GL}(N)} \text{trace}(\mathbf{T} \mathbf{P} \mathbf{T}^* + \mathbf{T}^{*-1} \mathbf{Q} \mathbf{T}^{-1})$$

└ Balancing

└ Variational Interpretation

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- For the optimal transformation \mathbf{T}_{bal} , the function takes the value

$$f(\mathbf{T}_{\text{bal}}) = 2\text{tr}(\mathbf{\Sigma}) = 2 \sum_{i=1}^N \sigma_i$$

where $\{\sigma_i\}_{i=1}^N$ are the Hankel singular values

- Balanced Truncation Method

- Block Partitioning of the System

- Applying the change of variable $\tilde{\mathbf{w}} = \mathbf{T}_{\text{bal}}\mathbf{w}$ transforms the given dynamical system into $(\tilde{\mathbf{A}}, \tilde{\mathbf{B}}, \tilde{\mathbf{C}})$ where

$$\tilde{\mathbf{A}} = \mathbf{T}_{\text{bal}}\mathbf{A}\mathbf{T}_{\text{bal}}^{-1}, \quad \tilde{\mathbf{B}} = \mathbf{T}_{\text{bal}}\mathbf{B}, \quad \tilde{\mathbf{C}} = \mathbf{C}\mathbf{T}_{\text{bal}}^{-1}$$

- Let $1 \leq k \leq N$; the system can be partitioned in blocks as

$$\tilde{\mathbf{A}} = \begin{bmatrix} \tilde{\mathbf{A}}_{11} & \tilde{\mathbf{A}}_{12} \\ \tilde{\mathbf{A}}_{21} & \tilde{\mathbf{A}}_{22} \end{bmatrix}, \quad \tilde{\mathbf{B}} = \begin{bmatrix} \tilde{\mathbf{B}}_1 \\ \tilde{\mathbf{B}}_2 \end{bmatrix}, \quad \tilde{\mathbf{C}} = \begin{bmatrix} \tilde{\mathbf{C}}_1 & \tilde{\mathbf{C}}_2 \end{bmatrix}$$

- The subscripts 1 and 2 denote the dimensions k and $N - k$, respectively

- The blocks with the subscript 1 correspond to the most observable and reachable states

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- Then, the following lower-dimensional model of size k

$$(\mathbf{A}_r, \mathbf{B}_r, \mathbf{C}_r) = (\tilde{\mathbf{A}}_{11}, \tilde{\mathbf{B}}_1, \tilde{\mathbf{C}}_1) \in \mathbb{R}^{k \times k} \times \mathbb{R}^{k \times in} \times \mathbb{R}^{q \times k}$$

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- The **left and right** ROBs are

$$\mathbf{W} = \mathbf{T}_{\text{bal}}^*(:, 1:k) \quad \text{and} \quad \mathbf{V} = \mathbf{S}_{\text{bal}}(:, 1:k), \quad \text{respectively,}$$

where $\mathbf{S}_{\text{bal}} = \mathbf{T}_{\text{bal}}^{-1}$

Definition (The Hardy space \mathcal{H}_∞)

The \mathcal{H}_∞ -norm associated with a system characterized by a transfer function $\mathbf{G}(\cdot)$ is defined as

$$\|\mathbf{G}\|_{\mathcal{H}_\infty} = \sup_{z \in \mathbb{C}_+} \|\mathbf{G}(z)\|_\infty = \sup_{z \in \mathbb{C}_+} \sigma_{\max}(\mathbf{G}(z))$$

where $z \in \mathbb{C}_+$ if $z \in \mathbb{C}$ and $\Im m(z) > 0$.

Proposition

$$(i) \quad \|\mathbf{G}\|_{\mathcal{H}_\infty} = \sup_{\omega \in \mathbb{R}} \|\mathbf{G}(i\omega)\|_\infty = \sup_{\omega \in \mathbb{R}} \sigma_{\max}(\mathbf{G}(i\omega))$$

$$(ii) \quad \|\mathbf{G}\|_{\mathcal{H}_\infty} = \sup_{\mathbf{u} \neq \mathbf{0}} \frac{\|\mathbf{y}(\cdot)\|_2}{\|\mathbf{u}(\cdot)\|_2} = \sup_{\mathbf{u} \neq \mathbf{0}} \sqrt{\frac{\int_0^\infty \|\mathbf{y}(t)\|_2^2 dt}{\int_0^\infty \|\mathbf{u}(t)\|_2^2 dt}}$$

- The \mathcal{H}_∞ norm of the error between the HDM- and PROM-based solutions will be used as an error criterion

Theorem (Error Bounds)

The BT procedure yields the following error bound for the output of interest.

Let $\{\tilde{\sigma}_i\}_{i=1}^{N_{SV}} \subseteq \{\sigma_i\}_{i=1}^N$ denote the distinct Hankel singular values of the system and $\{\tilde{\sigma}_i\}_{i=N_k+1}^{N_{SV}}$ the ones that have been truncated. Then

$$\|\mathbf{y}(\cdot) - \mathbf{y}_r(\cdot)\|_2 \leq 2 \sum_{i=N_k+1}^{N_{SV}} \tilde{\sigma}_i \|\mathbf{u}(\cdot)\|_2$$

Equivalently, the above result can be written in terms of the \mathcal{H}_∞ -norm of the system error as follows

$$\|\mathbf{G}(\cdot) - \mathbf{G}_r(\cdot)\|_{\mathcal{H}_\infty} \leq 2 \sum_{i=N_k+1}^{N_{SV}} \tilde{\sigma}_i$$

where \mathbf{G} and \mathbf{G}_r are the full- and reduced-order transfer functions. Equality holds when $\tilde{\sigma}_{N_k+1} = \tilde{\sigma}_{N_{SV}}$ (all truncated singular values are equal).

Proof. The proof proceeds in 3 steps:

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- 2 Show that if all truncated singular values are equal to σ , then

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- 3 Use this result to show that in the general case

$$\|\mathbf{E}(\cdot)\|_{\mathcal{H}_\infty} \leq 2 \sum_{i=N_k+1}^{n_{SV}} \tilde{\sigma}_i$$

Theorem (Stability Preservation)

Consider $(\mathbf{A}_r, \mathbf{B}_r, \mathbf{C}_r) = (\tilde{\mathbf{A}}_{11}, \tilde{\mathbf{B}}_1, \tilde{\mathbf{C}}_1)$, a PROM obtained by BT. Then

(i) $\mathbf{A}_r = \tilde{\mathbf{A}}_{11}$ has no eigenvalues in the open right half plane

(ii) Furthermore, if the systems $(\tilde{\mathbf{A}}_{11}, \tilde{\mathbf{B}}_1, \tilde{\mathbf{C}}_1)$ and $(\tilde{\mathbf{A}}_{22}, \tilde{\mathbf{B}}_2, \tilde{\mathbf{C}}_2)$ have no Hankel singular values in common, \mathbf{A}_r has no eigenvalues on the imaginary axis

- Because of numerical stability issues, computing the transformations

$$\mathbf{T}_{\text{bal}} = \mathbf{\Sigma}^{\frac{1}{2}} \mathbf{K}^* \mathbf{U}^{-1}, \quad \mathbf{T}_{\text{bal}}^{-1} = \mathbf{U} \mathbf{K} \mathbf{\Sigma}^{-\frac{1}{2}}$$

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- 3 construct the matrices

$$\mathbf{T}_{\text{bal}} = \mathbf{\Sigma}^{-\frac{1}{2}} \mathbf{V}^* \mathbf{Z}^* \quad \text{and} \quad \mathbf{T}_{\text{bal}}^{-1} = \mathbf{U} \mathbf{W} \mathbf{\Sigma}^{-\frac{1}{2}}$$

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- Proof: Show that $\mathbf{\Sigma}$ is real-valued then compute $\mathbf{T}_{\text{bal}} \mathcal{P} \mathbf{T}_{\text{bal}}^*$ and $\mathbf{T}_{\text{bal}}^{*-1} \mathcal{Q} \mathbf{T}_{\text{bal}}^{-1}$ using the above SVD

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- BT leads to PROMs with quality and stability guarantees; however
 - the computation of a Gramian is intensive as it requires the solution of a Lyapunov equation ($\mathcal{O}(N^3)$ operations)
 - for this reason, BT is in general impractical for large systems – say $N \gtrsim 10^5$ (but monitor progress in the literature if interested)

Recall the theorem underlying the construction of a POD basis

Theorem

Let $\hat{\mathbf{K}} \in \mathbb{R}^{N \times N}$ be the real symmetric semi-definite positive matrix defined as

$$\hat{\mathbf{K}} = \int_0^{\mathcal{T}} \mathbf{w}(t)\mathbf{w}(t)^T dt$$

Let $\hat{\lambda}_1 \geq \hat{\lambda}_2 \geq \dots \geq \hat{\lambda}_N \geq 0$ denote the ordered eigenvalues of $\hat{\mathbf{K}}$ and let $\hat{\phi}_i \in \mathbb{R}^N$, $i = 1, \dots, N$, denote the associated eigenvectors

$$\hat{\mathbf{K}}\hat{\phi}_i = \hat{\lambda}_i\hat{\phi}_i, \quad i = 1, \dots, N.$$

The subspace $\hat{\mathcal{V}} = \text{range}(\hat{\mathbf{V}})$ of dimension k minimizing $J(\mathbf{\Pi}_{\mathbf{V}, \mathbf{V}})$ is the invariant subspace of $\hat{\mathbf{K}}$ associated with the eigenvalues $\hat{\lambda}_1, \dots, \hat{\lambda}_k$

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- Consequently, the reachability Gramian is

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└ Comparison with the POD Method

└ POD for an Impulse Response

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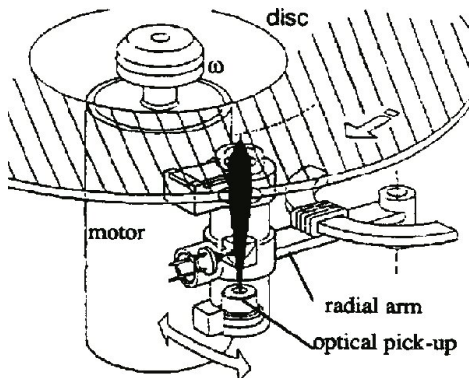
- Consequently, the reachability Gramian is

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- Unlike the BT method, the POD method does not take into account the observability Gramian to determine the PROM: therefore, every observable state may be truncated

Application

CD Player System (B. Salimbahrami and B. Lohmann, 2003)

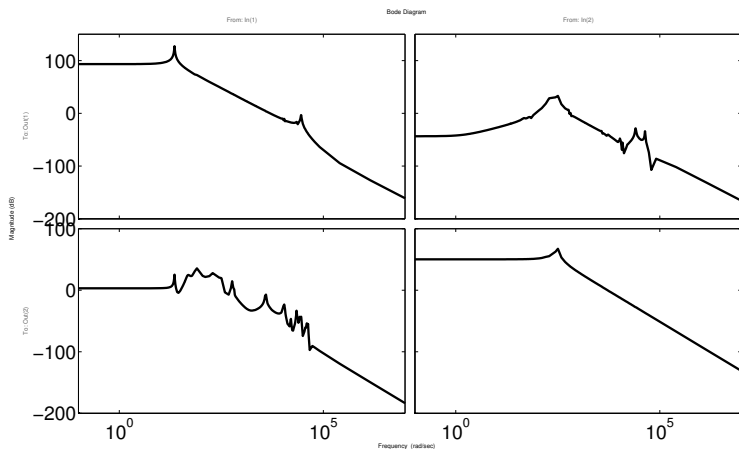


- Objective: model the position of the lens controlled by a swing arm
- System with $in = 2$ inputs and $q = 2$ outputs

Application

CD Player System (B. Salimbahrami and B. Lohmann, 2003)

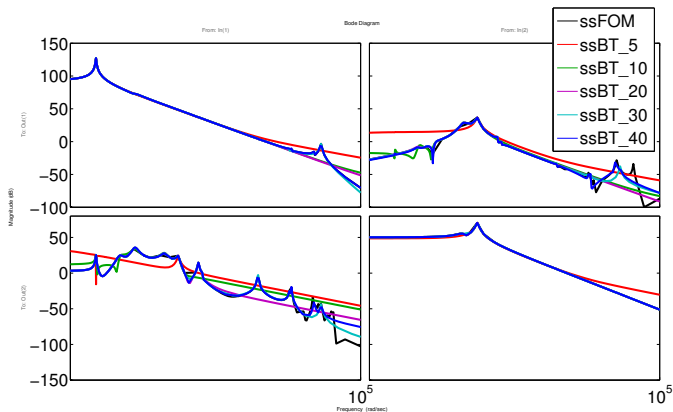
Bode plots associated with the HDM-based solution ($N = 120$): Each column represents one input and each row represents a different output



Application

CD Player System (B. Salimbahrami and B. Lohmann, 2003)

Bode plots associated with the PROM-based (BT) solution: Each column represents one input and each row represents a different output



└ Balanced POD Method

└ The Balanced POD Method

- The Balanced POD (BPOD) method generates two sets of snapshots: the standard POD solution snapshots; and the dual POD snapshots introduced below

$$\begin{aligned}\mathbf{S} &= [(j\omega_1 \mathbf{I} - \mathbf{A})^{-1} \mathbf{B} \cdots (j\omega_{N_{\text{snap}}} \mathbf{I} - \mathbf{A})^{-1} \mathbf{B}] \\ \mathbf{S}_{\text{dual}} &= [(-j\omega_1 \mathbf{I} - \mathbf{A}^*)^{-1} \mathbf{C}^* \cdots (-j\omega_{N_{\text{snap}}} \mathbf{I} - \mathbf{A}^*)^{-1} \mathbf{C}^*]\end{aligned}$$

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- Next, BPOD computes right and left ROBs as follows

$$\begin{aligned} \mathbf{S}_{\text{dual}}^T \mathbf{S} &= \mathbf{U} \mathbf{\Sigma} \mathbf{Z}^T \quad (\text{SVD}) \\ \mathbf{V} &= \mathbf{S} \mathbf{Z}_k \mathbf{\Sigma}_k^{-1/2} \\ \mathbf{W} &= \mathbf{S}_{\text{dual}} \mathbf{U}_k \mathbf{\Sigma}_k^{-1/2} \end{aligned}$$

where the subscript k designates the first k terms of the singular value decomposition

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where the subscript k designates the first k terms of the singular value decomposition

- If no truncation is performed, the result is equivalent to two-sided moment matching at $s_i \in \{\omega_1, \dots, \omega_{N_{\text{snap}}}\}$ (see later)

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- The POD method in the time domain is based solely on the reachability concept
- However, the BPOD method
 - introduces also the notion of observability in the construction of a PROM
 - is tractable for very large-scale systems
 - provides an approximation to the BT method
 - does not however guarantee in general the stability of the resulting PROM

- Supersonic Inlet Problem (part of the Oberwolfach Model Reduction Benchmark Collection repository)



$$\mathbf{E} \frac{d\mathbf{w}}{dt}(t) = \mathbf{A}\mathbf{w}(t) + \mathbf{B}\mathbf{u}(t)$$

$$\mathbf{y}(t) = \mathbf{C}\mathbf{w}(t)$$

- $N = 11\,370$ (2D Euler equations)
- $in = 1$ input (density disturbance of the inlet flow)
- $q = 1$ output (average Mach number at the diffuser throat)

└ Balanced POD Method

└ Application

- PMOR in the frequency domain using
 - POD
 - BPOD
- In both cases, same frequency sampling for the computation of solution snapshots

└ Balanced POD Method

└ Application

- PMOR in the frequency domain using
 - POD
 - BPOD
- In both cases, same frequency sampling for the computation of solution snapshots
- Plot of the magnitude of the relative error in the transfer function (within the sampled frequency interval) as a function of the dimension k of the constructed PROM

