

A new measure for distance to uncontrollability

Ivan Markovsky

University of Southampton

Outline

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Numerical rank of a matrix

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Algorithms

Linear time-invariant (LTI) systems

Let \mathcal{B} be LTI system of order n with m inputs and p outputs. Define

- in discrete-time: $(\sigma x)(t) := x(t+1)$ — **shift operator**
- in continuous-time: $\sigma x := dx/dt$ — **derivative operator**

Two common representations of LTI systems are

- **input/state/output representation**

$$\mathcal{B}_{i/s/o}(A, B, C, D) := \{ \text{col}(u, y) \mid \exists x, \sigma x = Ax + Bu, y = Cx + Du \}$$

- **input/output representation**

$$\mathcal{B}_{i/o}(P, Q) := \{ \text{col}(u, y) \mid P(\sigma)y = Q(\sigma)u \}$$

where $P \in \mathbb{R}^{p \times p}[\xi]$, $\det(P) \neq 0$ and $Q \in \mathbb{R}^{p \times m}[\xi]$.

Controllability

Definition \mathcal{B} is controllable if for all $w_1, w_2 \in \mathcal{B}$, $\exists w \in \mathcal{B}$, $\tau > 0$, such that $w_1(t) = w(t)$, for all $t < 0$ and $w_2(t) = w(t)$, for all $t \geq \tau$.

Think of w_1 as a given past traj. and w_2 as a desired future traj.

\mathcal{B} controllable \implies any given traj. can be steered to any desired trajectory

important condition for pole-placement, LQ, H_∞ , ... control, e.g.,

controllability \iff solvability of the state feedback pole-placement problem

Controllability test in terms of I/S/O representation

For numerically checking controllability of \mathcal{B} , we need to relate this property to the parameters of \mathcal{B} in a particular representation.

Consider an I/S/O representation $\mathcal{B} = \mathcal{B}_{i/s/o}(A, B, C, D)$.

A well known result is that

$\mathcal{B}_{i/s/o}$ is controllable iff $\mathcal{C} := \begin{bmatrix} A & BA & \cdots & BA^{n-1} \end{bmatrix}$ is full rank

\Rightarrow checking controllability is a **rank test problem** for a structured matrix, which is a nonlinear transformation of A, B

Distance to rank deficiency

In numerical linear algebra, **yes/no questions** (\mathcal{B} contr./uncontr.) are replaced by **quantitative measures** (distance of \mathcal{B} to uncontr.)

Checking whether \mathcal{C} is full rank is a yes/no question.

A corresponding quantitative measure is distance of \mathcal{C} to rank deficiency: **smallest $\|\Delta\mathcal{C}\|$, such that $\widehat{\mathcal{C}} := \mathcal{C} + \Delta\mathcal{C}$ is rank def.**

However for $\|\Delta\mathcal{C}\|$ to be a meaningful measure for distance to uncontr., $\widehat{\mathcal{C}}$ has to be a controllability matrix for some system $\widehat{\mathcal{B}}$.

$\Rightarrow \Delta\mathcal{C}$ should have the same structure as \mathcal{C} .

Controllability test in terms of I/O representation

Consider an I/O representation $\mathcal{B} = \mathcal{B}_{i/o}(P, Q)$.

A well known result is that

$\mathcal{B}_{i/o}(P, Q)$ is controllable if and only if P and Q are coprime.

\Rightarrow checking controllability is a **coprimeness test problem** for a pair of polynomial matrices.

Unstructured/structured low rank approximation

Consider a set of **structured matrices** \mathbb{M} and define

$$d_r(A) := \min_{\Delta A \in \mathbb{M}} \|\Delta A\| \quad \text{subject to} \quad A + \Delta A \text{ has rank } r.$$

With $\mathbb{M} = \mathbb{R}^{m \times n}$, $d_r(A)$ is **unstructured** distance to rank- r matrices.

In special cases, unstructured $d_r(A)$ can be computed from the SVD of A

$$A = U \text{diag}(\sigma_1, \dots, \sigma_{\min(m,n)}) V^T$$

- spectral norms: $d_r(A) = \sigma_{r+1}$
- Frobenius norm: $d_r(A) = \sqrt{\sigma_{r+1}^2 + \cdots + \sigma_{\min(m,n)}^2}$.

In general, $d_r(A)$ is difficult to compute.

Paige's distance to uncontrollability

C. C. Paige defined in

Properties of numerical algorithms related to computing controllability, IEEE-AC, vol. 26, 1981

the following measure for distance of \mathcal{B} to uncontrollability

$$d(A, B) := \min_{\hat{A}, \hat{B}} \left\| \begin{bmatrix} A & B \end{bmatrix} - \begin{bmatrix} \hat{A} & \hat{B} \end{bmatrix} \right\|_F$$

subject to $\mathcal{C}(\hat{A}, \hat{B})$ is rank deficient

many papers on computing $d(A, B)$ (98 citations in WoS)

However, $d(A, B)$ depends on the choice of the state space basis!

$\Rightarrow d(A, B)$ not a genuine property of the pair of systems $(\mathcal{B}, \hat{\mathcal{B}})$

Special case: I/O representation of a SISO system

Consider a **SISO system** \mathcal{B} with an input/output representation

$$p(\sigma)y = q(\sigma)u$$

With p monic, p, q are unique and

$$\text{dist}(\mathcal{B}, \hat{\mathcal{B}}) := \sqrt{\|p - \hat{p}\|_2^2 + \|q - \hat{q}\|_2^2}$$

becomes a property of the pair of systems $(\mathcal{B}, \hat{\mathcal{B}})$.

The problem of computing $d(\mathcal{B}_{i/o}(p, q))$ becomes

$$\min_{\hat{p}, \hat{q}} \left\| \begin{bmatrix} p \\ q \end{bmatrix} - \begin{bmatrix} \hat{p} \\ \hat{q} \end{bmatrix} \right\|_2 \quad \text{subject to} \quad \mathcal{B}_{i/o}(\hat{p}, \hat{q}) \in \overline{\mathcal{L}_{\text{ctrb}}} \quad (*)$$

More general definition

$$d(\mathcal{B}) := \min_{\hat{\mathcal{B}} \in \overline{\mathcal{L}_{\text{ctrb}}}} \text{dist}(\mathcal{B}, \hat{\mathcal{B}})$$

where

- $\overline{\mathcal{L}_{\text{ctrb}}}$ is the set of uncontrollable LTI systems
- $\text{dist}(\mathcal{B}, \hat{\mathcal{B}})$ is a measure for the distance from \mathcal{B} to $\hat{\mathcal{B}}$

Note: $d(A, B)$ is formally a special case of $d(\mathcal{B})$ with

$$\text{dist}(\mathcal{B}, \hat{\mathcal{B}}) = \left\| \begin{bmatrix} A & B \end{bmatrix} - \begin{bmatrix} \hat{A} & \hat{B} \end{bmatrix} \right\|_F \quad (\text{Paige})$$

however given \mathcal{B} and $\hat{\mathcal{B}}$, A, B and \hat{A}, \hat{B} are not uniquely defined

\Rightarrow (Paige) is not well defined

Representations of $\mathcal{B}_{i/o}(\hat{p}, \hat{q}) \in \overline{\mathcal{L}_{\text{ctrb}}}$

$$\mathcal{B}_{i/o}(\hat{p}, \hat{q}) \in \overline{\mathcal{L}_{\text{ctrb}}} \iff \hat{p}, \hat{q} \text{ have common divisor } c, \deg(c) \geq 1$$

$$\iff \exists u, v \text{ such that } \begin{aligned} \hat{p}(\xi) &= u(\xi)c(\xi) \\ \hat{q}(\xi) &= v(\xi)c(\xi) \end{aligned}$$

$$\iff \exists u, v \text{ such that } \hat{p}(\xi)v(\xi) = \hat{q}(\xi)u(\xi)$$

$$\iff \exists u, v \text{ such that } S(\hat{p}, \hat{q}) \begin{bmatrix} v \\ -u \end{bmatrix} = 0$$

$$\iff S(\hat{p}, \hat{q}) \text{ is rank deficient}$$

Equivalent problem to (*)

$\widehat{\mathcal{B}} \in \overline{\mathcal{L}_{\text{ctrb}}}$ is equivalent to rank deficiency of the Sylvester matrix

$$S(\widehat{p}, \widehat{q}) := \begin{bmatrix} \widehat{p}_0 & & & \widehat{q}_0 & & \\ \widehat{p}_1 & \widehat{p}_0 & & \widehat{q}_1 & \widehat{q}_0 & \\ \vdots & \widehat{p}_1 & \ddots & \vdots & \widehat{q}_1 & \ddots \\ \widehat{p}_n & \vdots & \ddots & \widehat{p}_0 & \widehat{q}_n & \vdots & \ddots & \widehat{q}_0 \\ & \widehat{p}_n & & \widehat{p}_1 & \widehat{q}_n & & \ddots & \widehat{q}_1 \\ & & \ddots & \vdots & & & \ddots & \vdots \\ & & & \widehat{p}_n & & & & \widehat{q}_n \end{bmatrix}$$

problem (*) is a **Sylvester structured low-rank approximation**

$$\min_{\widehat{p}, \widehat{q}, w} \left\| \begin{bmatrix} p \\ q \end{bmatrix} - \begin{bmatrix} \widehat{p} \\ \widehat{q} \end{bmatrix} \right\|_2 \quad \text{subject to} \quad S(\widehat{p}, \widehat{q}) \begin{bmatrix} w \\ 1 \end{bmatrix} = 0 \quad (**)$$

Local optimization based algorithm

Theorem $d(\mathcal{B}_{\text{i/o}}(p, q))$ is equal to

$$\min_{c \in \mathbb{R}} \text{trace} \left(\begin{bmatrix} p & q \end{bmatrix}^\top \left(I - T([1]) (T^\top([1]) T([1]))^{-1} T^\top([1]) \right) \begin{bmatrix} p & q \end{bmatrix} \right).$$

Notes:

- \widehat{p} , \widehat{q} , u , v and the constraint are eliminated from (***)
- nonconvex nonlinear least squares problem
- solved numerically using local optimization methods
- the optimization variable is a scalar
- cost function evaluations: solve a structured LS problem
- exploiting structure, comput. complexity per iteration $O(n)$

Another equivalent problem to (*)

$$a(\xi)b(\xi) \iff \underbrace{\begin{bmatrix} a_0 & & & & \\ a_1 & a_0 & & & \\ \vdots & a_1 & \ddots & & \\ a_{n_a} & \vdots & \ddots & a_0 & \\ & a_{n_a} & & a_1 & \\ & & \ddots & \vdots & \\ & & & a_{n_a} \end{bmatrix}}_{T(a)} b \iff T(b)a$$

From $\mathcal{B}_{\text{i/o}}(\widehat{p}, \widehat{q}) \in \overline{\mathcal{L}_{\text{ctrb}}} \iff \exists u, v, \text{ s.t. } \begin{cases} \widehat{p}(\xi) = u(\xi)c(\xi) \\ \widehat{q}(\xi) = v(\xi)c(\xi) \end{cases}$,
we have

$$\min_{\widehat{p}, \widehat{q}, u, v, c} \left\| \begin{bmatrix} p \\ q \end{bmatrix} - \begin{bmatrix} \widehat{p} \\ \widehat{q} \end{bmatrix} \right\|_2 \quad \text{subject to} \quad \begin{bmatrix} \widehat{p} \\ \widehat{q} \end{bmatrix} = \begin{bmatrix} T(u) \\ T(v) \end{bmatrix} c \quad (***)$$

Proof

Rewrite the constraint of (***) as follows

$$\begin{bmatrix} \widehat{p} \\ \widehat{q} \end{bmatrix} = \begin{bmatrix} T(u) \\ T(v) \end{bmatrix} c \iff \begin{bmatrix} \widehat{p} & \widehat{q} \end{bmatrix} = T(c) \begin{bmatrix} u & v \end{bmatrix}$$

so that

$$\left\| \begin{bmatrix} p \\ q \end{bmatrix} - \begin{bmatrix} \widehat{p} \\ \widehat{q} \end{bmatrix} \right\|_2 = \left\| \begin{bmatrix} p & q \end{bmatrix} - T(c) \begin{bmatrix} u & v \end{bmatrix} \right\|_F$$

(***) becomes an ordinary least-squares problem in u, v

\rightsquigarrow closed form expression in c

Suboptimal initial approximations

can be computed from unstructured low rank approx. (SVD) of

1. Sylvester matrix $S(p, q)$
2. Bezout matrix $B(p, q)$
3. Hankel matrix $H(h)$
4. Balanced model reduction

$$B(p, q) := \begin{bmatrix} p_1 & \cdots & p_n \\ \vdots & \ddots & \vdots \\ p_n & & 0 \end{bmatrix} \begin{bmatrix} q_0 & \cdots & q_{n-1} \\ & \ddots & \vdots \\ 0 & & q_{n-1} \end{bmatrix} - \begin{bmatrix} q_1 & \cdots & q_n \\ \vdots & \ddots & \vdots \\ q_n & & 0 \end{bmatrix} \begin{bmatrix} p_0 & \cdots & p_{n-1} \\ & \ddots & \vdots \\ 0 & & p_{n-1} \end{bmatrix}$$

$$H(h) := \begin{bmatrix} h_1 & h_2 & \cdots & h_n \\ h_2 & h_3 & \ddots & h_{n+1} \\ \vdots & \ddots & \ddots & \vdots \\ h_n & h_{n+1} & \cdots & h_{2n} \end{bmatrix}, \quad \frac{q(z)}{p(z)} = \sum_{t=0}^{\infty} h_t z^{-t-1}$$

Simulation example

Given $\mathcal{B}_{i/o}(p, q)$, where

$$p(\xi) = 0.058 + 0.684\xi + 2.745\xi^2 + 4.751\xi^3 + 3.622\xi^4 + 1.000\xi^5$$

$$q(\xi) = -0.134 - 1.408\xi - 5.149\xi^2 - 8.381\xi^3 - 6.092\xi^4 - 1.604\xi^5$$

we compute initial approximation $\mathcal{B}_{i/o}(\hat{p}, \hat{q})$ by one of the suboptimal methods and a locally optimal solution $\mathcal{B}_{i/o}(\hat{p}^*, \hat{q}^*)$

Method	$d(\mathcal{B}_{i/o}(\hat{p}, \hat{q}))$	$d(\mathcal{B}_{i/o}(\hat{p}^*, \hat{q}^*))$	# iter.	# cost fun. eval.
Sylvester	0.0109	0.0021	4	20
Bezout	0.0041	0.0007	3	12
Hankel	0.0026	0.0007	3	12
BMR	0.0051	0.0007	5	46

(BMR — Balanced model reduction)

Using the Sylvester matrix

$$\mathcal{B}_{i/o}(p, q) \in \overline{\mathcal{L}_{\text{ctrb}}} \iff \exists u, v \text{ such that } S(p, q) \begin{bmatrix} v \\ -u \end{bmatrix} = 0$$

1. Compute the SVD of $S(p, q) \rightsquigarrow$ approximate u and v .
2. Solve the LS problem $\begin{bmatrix} p \\ q \end{bmatrix} = \begin{bmatrix} T(u) \\ T(v) \end{bmatrix} c$.
3. Define

$$\hat{p}(\xi) = u(\xi) a_{\text{ls}}(\xi) \quad \text{and} \quad \hat{q}(\xi) = v(\xi) a_{\text{ls}}(\xi).$$

Then

$$d(\mathcal{B}_{i/o}(p, q)) \leq \|\text{col}(p, q) - \text{col}(\hat{p}, \hat{q})\|_2.$$

Conclusions

- **motivation:** replace the statement “ \mathcal{B} contr./uncontr.” with a quantitative one “distance of \mathcal{B} to uncontrollability”
- the definition invariably considered in the literature is **not representation invariant**
- **behavioral measure:** $d(\mathcal{B}) := \min_{\hat{\mathcal{B}} \in \overline{\mathcal{L}_{\text{ctrb}}}} \text{dist}(\mathcal{B}, \hat{\mathcal{B}})$
- in the **SISO case**, $d(\mathcal{B})$ can be defined in terms of the normalized I/O representation $p(\sigma)y = q(\sigma)u$
- the **computation of $d(\mathcal{B})$** leads to a nonlinear least squares problem, which cost function evaluation is $O(n)$
- **SVD upper bounds**, based on the Sylvester, Bezout, Hankel matrices, and balanced model reduction