



MAX PLANCK INSTITUTE
FOR DYNAMICS OF COMPLEX
TECHNICAL SYSTEMS
MAGDEBURG



COMPUTATIONAL METHODS IN
SYSTEMS AND CONTROL THEORY

MORscore – Comparability of Model Reduction Algorithms

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Identification, Simulation and Control of Complex Dynamical Systems from Data
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Supported by:

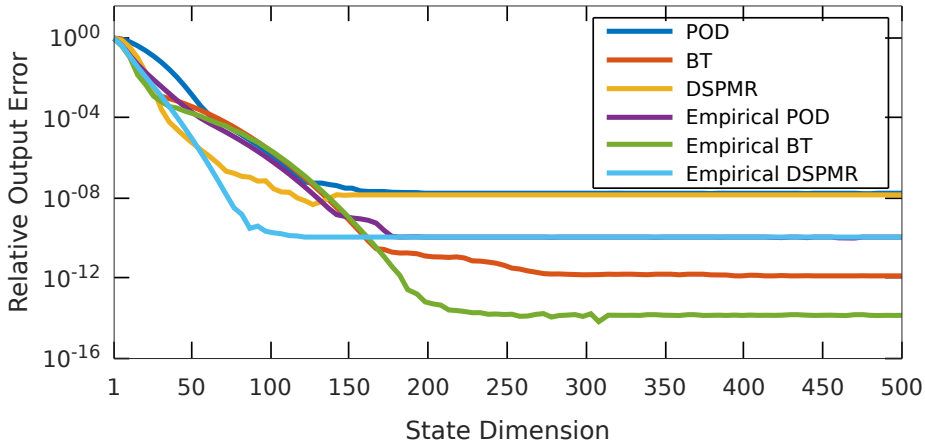


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When we (should) compare MOR algorithms:

- New Methods (qualitative)
- MORwiki (quantitative)
- Software updates (regressions)
- Reviews (verification)
- Collaborations (recommendations)

But: What means better?

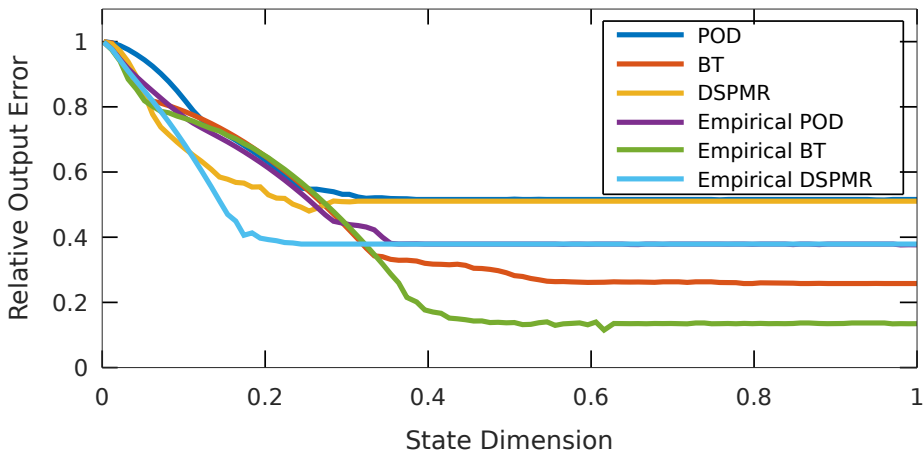


- How should we compare MOR algorithms?
→ It is a multi-objective problem.
- Can we assign a performance index to a method?
→ Like in computer benchmarking.
- Can we come up with a readable (ordered) score?
→ Humans should be able to easily comprehend.

1. Select maximum reduced order: $n_{\max} := 100$
2. Normalize orders: $1, 2, \dots, 100 \rightarrow \frac{1}{100}, \frac{2}{100}, \dots, \frac{100}{100}$
3. Normalize errors (via \log_{10}): $10^0, 10^{-1}, \dots, 10^{-16} \rightarrow 1, \frac{15}{16}, \dots, \frac{1}{16}$

Now we have:

- normalized orders $n_i \in (0, 1] \subset \mathbb{Q}$, and
- normalized errors $\varepsilon_i \in [0, 1] \subset \mathbb{R}$.



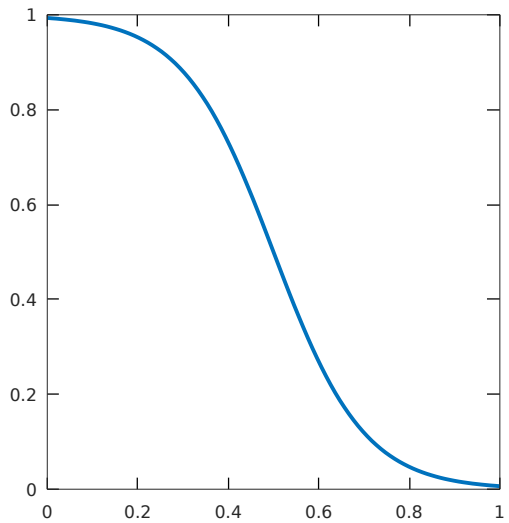
Normalized Error Mapping:

$$\varepsilon_{\text{MOR}} : (0, 1] \subset \mathbb{Q} \rightarrow [0, 1] \subset \mathbb{R}$$

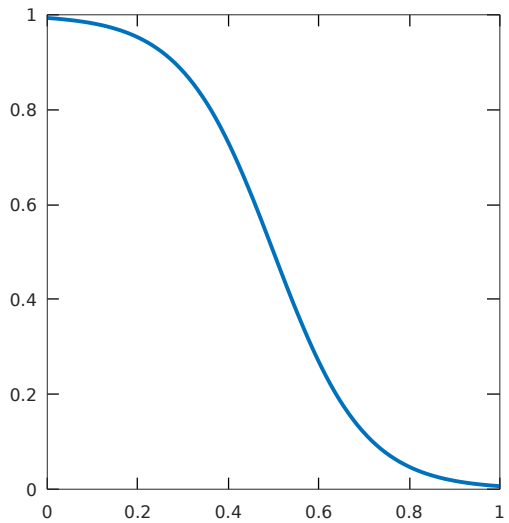
Normalized Error Graph:

$$(n_i, \varepsilon_i)$$

- Maps a normalized reduced order to a normalized relative log-error.
- Typically gathered through empirical testing.
- Alternatively error bounds or indicators could be used.



- Errors start near one,
- decay with some speed,
- and then flatten out.



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- decay with some speed,
- and then flatten out.

Like a sigmoid function!

Sigmoid (Logistic Function):

$$s(x) = 1 - \frac{L}{1 + e^{-k(x-x_0)}}$$

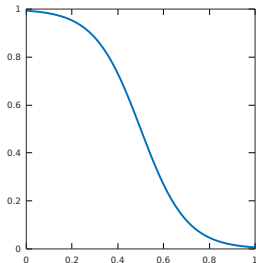
Parameters:

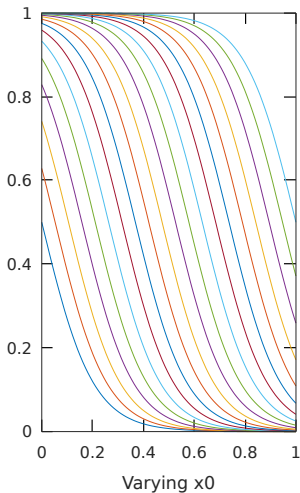
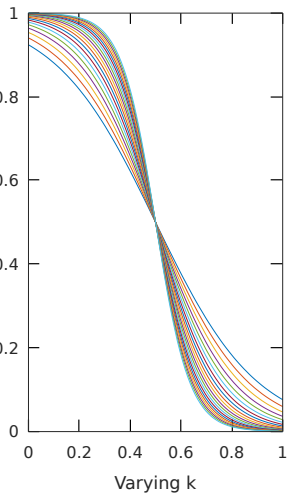
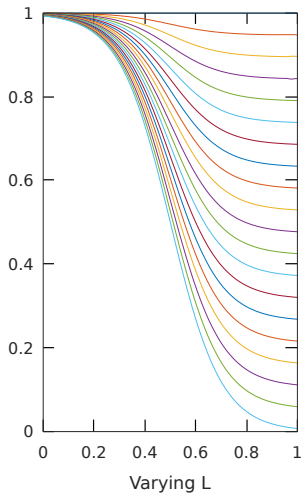
- L Minimum (higher means better)
- k Steepness (higher means better)
- x_0 Midpoint (higher means better)

Score is least-squares fit to error graph.

Comments:

- How to compare two scores? In a norm?
- How should a norm be weighted?
- Is it useful?





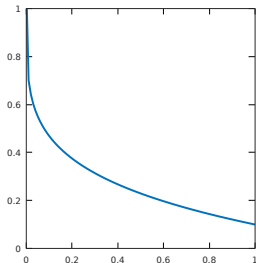
Monomial:

$$m(x) = 1 - (x^d + c)$$

Parameters:

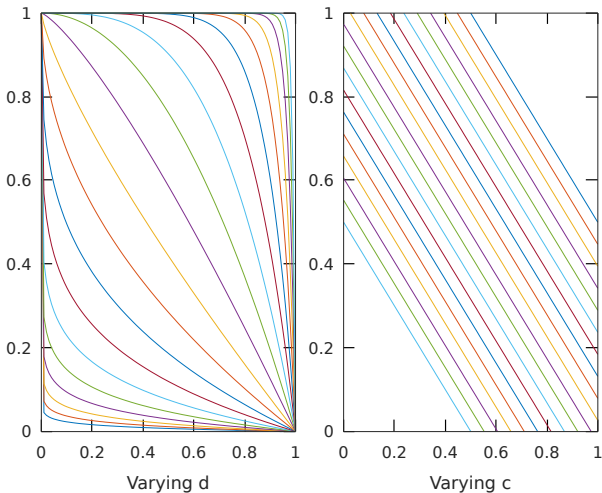
- $d > 0$ Degree (lower means better)
- c Shift (lower means better)

Score is least-squares fit to error graph.



Comments:

- How to compare two scores?
- What means a different shift?
- Is it useful?



Trapezoid Rule:

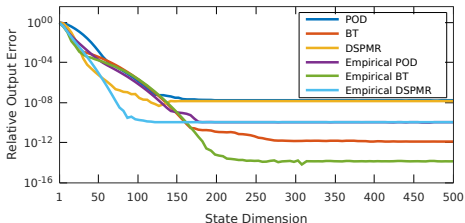
$$A = 1 - \sum_{k=1}^N \frac{\Delta x}{2} (y_{k-1} + y_k)$$

Candidate:

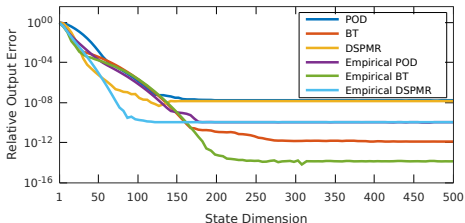
- $A = \text{Area above curve}$

Comments:

- One score!
- Cheap computation.
- See also: [Curtis, Mitchell, Overton'17]



	Sigmoid (L, k, x_0)	Monomial (d, c)	Trapezoid
POD	(0.48, 20.85, 0.14)	(0.15, -0.45)	0.42
BT	(0.74, 11.30, 0.20)	(0.33, -0.16)	0.59
DSPMR	(0.49, 30.32, 0.08)	(0.11, -0.45)	0.45
Empirical POD	(0.63, 13.83, 0.16)	(0.22, -0.30)	0.53
Empirical BT	(0.88, 10.70, 0.23)	(0.45, -0.01)	0.68
Empirical DSPMR	(0.62, 29.40, 0.10)	(0.16, -0.31)	0.56



	Sigmoid (L, k, x_0)	Monomial (d, c)	Trapezoid
POD	(0.48, 10.40, 0.27)	(0.24, -0.46)	0.35
BT	(0.75, 5.53, 0.40)	(0.47, -0.24)	0.44
DSPMR	(0.49, 15.15, 0.16)	(0.18, -0.44)	0.41
Empirical POD	(0.64, 6.51, 0.32)	(0.34, -0.32)	0.43
Empirical BT	(1.05, 4.18, 0.55)	(0.92, -0.03)	0.49
Empirical DSPMR	(0.62, 14.46, 0.20)	(0.27, -0.29)	0.50

MORscore := The area fraction of the normalized log-error graph.

Questions:

- How to select maximum reduced order?
- Would this be useful to you?
- Should it be used in the MORwiki?
- How to handle unstable ROMs?
- Do you have other ideas for scores?