

Model Order Reduction or How to make everything as simple as possible but not simpler

Chahlaoui, Younes

2011

MIMS EPrint: 2011.51

## Manchester Institute for Mathematical Sciences School of Mathematics

The University of Manchester

Reports available from: http://eprints.maths.manchester.ac.uk/ And by contacting: The MIMS Secretary School of Mathematics The University of Manchester Manchester, M13 9PL, UK

ISSN 1749-9097

MAMERN11: 4<sup>th</sup> International Conference on Approximation Methods and Numerical Modelling in Environment and Natural Resources Saidia (Morocco), May 23-26, 2011

## Model Order Reduction or How to make everything as simple as possible but not simpler

## Younès Chahlaoui

Centre for Interdisciplinary Computational and Dynamical Analysis The University of Manchester, UK Younes.Chahlaoui@manchester.ac.uk

**Keywords:** Model order reduction, large-scale system, low-dimensional approximations.

**Abstract.** Large complex mathematical models are regularly used for simulation and prediction. However, in control design it is common practice to work with as simple models as possible, because they are easier to analyse and evaluate.

There is a strong need for methods and tools that can take a complex model and deduce simple models for various purposes such as control design. A simple but good model captures much knowledge. It points out the basic properties and can give good insight about the process.

For simple linear time-invariant models there is a well-established theory and commercially available tools for design of controllers with given specifications. Real experiments or simulations using more complex models are then used to verify that the designed controller really works well. For nonlinear models the methods are much less developed. It is simple to derive a linearization on symbolic form from a nonlinear model. It is much more difficult to give explicit expressions for stationary operating points since these calculations involve nonlinear equation systems.

The main idea in model reduction is that a high-dimensional state vector

1

is actually belonging to a low-dimensional subspace [1, 2, 4]. Provided that the low-rank subspace is known, the original model can be projected on it to obtain a required low-dimensional approximation [3]. The goal of every model reduction method is to find such a low-dimensional subspace.

In this talk I will introduce model reduction and I will overview some of the most used methods.

## REFERENCES

- [1] A.C. Antoulas, *Approximation of Large-Scale Dynamical Systems*, SIAM, 2005.
- [2] P. Benner, V. Mehrmann, and D.C. Sorensen, *Dimension Reduction of Large-Scale Systems*. Lecture Notes in Computational Science and Engineering, Vol. 45, Springer-Verlag, Berlin/Heidelberg, 2005.
- [3] Y. Chahlaoui, *Low-rank approximation and model reduction*. PhD thesis number 14/2003, UCLouvain, 2003.
- [4] W. Schilders, H. van der Vorst, and J. Rommes, *Model order reduction: Theory, research aspects and applications*. Mathematics in Industry 13. Berlin: Springer, 2008.