MODEL ORDER REDUCTION



OUTLINE

- Need for model order reduction (MOR)
- Methodology for MOR
- Classical approaches
 - Supercapacitor
- Data driven approaches
 - Design of experiment
 - Linear approach: Batteries
 - Nonlinear SISO / MISO approach: Power converter
 - Nonlinear MIMO approach: Fuel cell
- Summary of findings

MODEL ORDER REDUCTION OVERVIEW

• MOR is required to reduce computational complexity of model yet retain sufficient accuracy for a specific purpose, i.e. control, diagnosis, prognosis



METHODOLOGY

- Models for purpose:
 - Control
 - Diagnosis
 - Prognosis
- Approaches for MOR:
 - Classical
 - Based on mathematical manipulation of system equations

Data-driven

- Models derived from data collected from complex models or hardware
- Include System Identification and Machine Learning methods
- All approaches retain dominant modes whilst discarding modes with low contribution to system dynamics



METHODOLOGY



CASE STUDY: SUPERCAPACITOR

- Single input single output (SISO) nonlinear model
- 60th order model used as baseline for model order reduction
- Truncation without model order reduction methods



CASE STUDY: SUPERCAPACITOR



DATA DRIVEN APPROACHES

- An experimental process, whereby making use of available input-output data and a priori knowledge, one aims to mathematically describe causalities that govern behaviour of system
- Different approaches to modelling based on a priori knowledge
 - White box
 - Black box
 - Grey box





CASE STUDY: BATTERY

1. Data acquisition

- Voltage and SOC (outputs) responses to current input
- 36 short (80 seconds) data sets starting at different SOC (positive current input – charge mode)
- Experiment repeated for negative current (discharging)
- 2. Obtained set of 144 LTI models using simplified refined instrumental variable (SRIVC) method
 - Current to SOC models
 - Current to voltage models
- **3. Assumption:** low order model can have linear structure, where parameters depend on SOC and sign of current

CASE STUDY: BATTERY



CASE STUDY: BUCK-BOOST CONVERTER

Model structure : 10 single input - single output (SISO) Hammerstein models with 4th order polynomial static nonlinearity

- Inputs: Input voltage and duty cycle
- Output: Output current
- Model order: 1
- $R_{\tau}^2 = 99.83 \%$





CASE STUDY: BUCK-BOOST CONVERTER

Model structure : Bilinear multiple inputs – single output (MISO)

- Inputs: Input voltage and duty cycle
- Output: Output current
- Model order: 5
- $R_T^2 = 93.7 \%$
- a 5
- b 2
- Bilinear term 1



CASE STUDY: FUEL CELL STACK



Model structure : MIMO NARX

- Inputs: stack current, compressor voltage
- Outputs: stack voltage, oxygen excess, net power
- Logarithmic type nonlinearity on the input
- Model order: 15
- Oxygen excess R_{τ}^2 = 96.8 %
- Stack voltage $R_T^2 = 94.2 \%$
- Net power $R_{T}^{2} = 99.3 \%$

J. Pukrushpan, A. Stefanopoulou, and H. Peng, "Control of fuel cell breathing," IEEE Contr. Syst. Mag., vol. 24, no. 2, pp. 30–46, Apr. 2004.





SUMMARY OF FINDINGS

- MOR effectively retains fidelity of high order model whilst reducing the model order
- Data driven approaches are effective for reduced order modelling
- Purpose of model and a priori information determines the modelling method

- Outline of methodology for model order reduction
 - Control
 - Diagnosis
 - Prognosis
- Guidelines for MOR
- Methodology is robust for multiple case studies