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An automated surrogate modelling approach for the uncertainty quantification of dynamical systems

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1 Problem Statement

The challenge: Build a surrogate $\tilde{\mathcal{M}}$ that emulates the response of a complex time-dependent system \mathcal{M} over extended time periods:

$$y(t) = \mathcal{M}(\mathbf{x}(\mathcal{T} \leq t)) \approx \tilde{\mathcal{M}}(\mathbf{x}(\mathcal{T} \leq t))$$

- Discretized time axis $\mathcal{T} = \{0, \delta t, 2\delta t, \dots, (N-1)\delta t\}$
- System response $y : \mathcal{T} \rightarrow \mathbb{R}$
- Possibly high-dimensional exogenous excitation $\mathbf{x} : \mathcal{T} \rightarrow \mathbb{R}^M$

Our approach: Automated incremental construction of an exogenous input manifold suitable for the construction of autoregressive surrogates

2 Autoregressive Modelling

Nonlinear AutoRegressive with eXogenous input (NARX) models take advantage of the temporal coherence of the output and exogenous inputs:

$$y(t) = \tilde{\mathcal{M}}(\varphi(t), \mathbf{c})$$

$$\varphi(t) = \{y(t - \ell_1^y), \dots, y(t - \ell_{n_y}^y), x_1(t - \ell_1^{x_1}), \dots, x_1(t - \ell_{n_{x_1}}^{x_1}), \dots, x_{M_x}(t - \ell_1^{x_{M_x}}), \dots, x_{M_x}(t - \ell_{n_{x_{M_x}}}^{x_{M_x}})\}$$

- Model parameters \mathbf{c}
- $\ell_i^y \in \{\delta t, \dots, (N-1)\delta t\}$
- $\ell_i^{x_j} \in \{0, \delta t, \dots, (N-1)\delta t\}$

3 Manifold NARX

Exogenous input manifold

$$\tilde{\mathcal{M}} : \zeta(\mathcal{T} \leq t) \rightarrow y(t)$$

- Manifold ζ enables simpler mapping to \mathbf{y}
- Incorporates prior knowledge about the system \mathcal{M}
- Consists of auxiliary quantities \mathbf{z}_i

Auxiliary quantities

$$\begin{aligned} \mathbf{z}_1(t) &= \mathcal{F}_1(\mathbf{x}(\mathcal{T} \leq t), \mathbf{z}_1(\mathcal{T} < t)) \\ \mathbf{z}_2(t) &= \mathcal{F}_2(\mathbf{x}(\mathcal{T} \leq t), \mathbf{z}_1(\mathcal{T} \leq t), \mathbf{z}_2(\mathcal{T} < t)) \\ &\vdots \\ \mathbf{z}_i(t) &= \mathcal{F}_i(\mathbf{x}(\mathcal{T} \leq t), \mathbf{z}_1(\mathcal{T} \leq t), \dots, \mathbf{z}_{i-1}(\mathcal{T} \leq t), \mathbf{z}_i(\mathcal{T} < t)) \end{aligned}$$

- Transform \mathcal{F}_i can be an ARX model
- Auxiliary quantities can depend on each other
- Examples are control system outputs, moving averages, integrals or derivatives

Automatic recursive selection of auxiliary quantities

```

Function SelectFeatures( $\tilde{\mathbf{x}}, \mathbf{z}, \mathbf{y}$ )
 $\zeta \leftarrow \{\}, \tilde{\mathbf{y}} \leftarrow \mathbf{y}$ 
while  $\tilde{\mathbf{x}} \neq \{\}$  or  $\mathbf{z} \neq \{\}$  do
   $\rho \leftarrow \text{Correlate}(\{\tilde{\mathbf{x}}, \mathbf{z}\}, \tilde{\mathbf{y}})$ 
  if  $\max(|\rho|) < \theta$  then
    break
   $\zeta_i \leftarrow \arg \max(|\rho|)$ 
  if  $\zeta_i \in \tilde{\mathbf{x}}$  then
     $\zeta \leftarrow \{\zeta, \zeta_i\}$ 
     $\tilde{\mathbf{x}} \leftarrow \tilde{\mathbf{x}} \setminus \zeta_i$ 
  else if  $\zeta_i \in \mathbf{z}$  then
     $\zeta' \leftarrow \text{SelectFeatures}(\{\tilde{\mathbf{x}}, \zeta\}, \mathbf{z} \setminus \zeta_i, \zeta_i)$ 
     $\tilde{\mathbf{x}} \leftarrow \{\tilde{\mathbf{x}}, \mathbf{z} \cap \zeta'\}$ 
     $\mathbf{z} \leftarrow \mathbf{z} \setminus \zeta_i$ 
     $\zeta \leftarrow \{\zeta, \zeta_i\}$ 
     $\mathbf{z} \leftarrow \mathbf{z} \setminus \zeta_i$ 
   $\tilde{\mathbf{y}} \leftarrow \mathbf{y} - \tilde{\mathcal{M}}(\zeta)$ 
return  $\zeta$ 

Function Correlate( $\mathbf{x}, \mathbf{y}$ )
 $\rho \leftarrow \{\}$ 
for  $x_i \leftarrow x_1, \dots, x_M$  do
   $\rho \leftarrow \{\rho, \text{KendallTau}(x_i, \mathbf{y})\}$ 
return  $\rho$ 

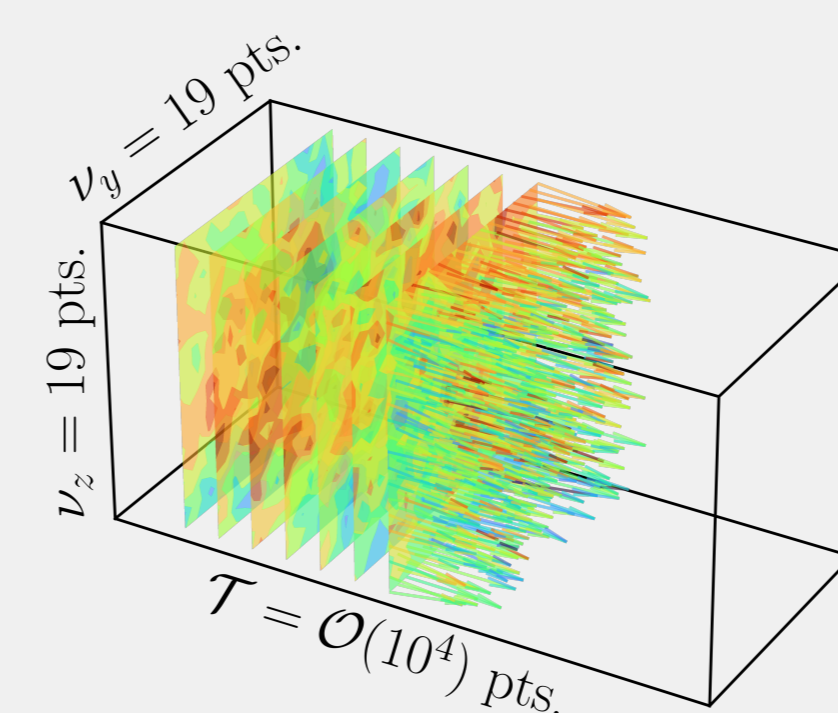
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- $\tilde{\mathbf{x}}$: available features
- \mathbf{z} : yet unavailable features
- $\tilde{\mathbf{y}}$: residuals
- ρ : measure of association
- θ : critical correlation value
- $\tilde{\mathcal{M}}$: proxy surrogate

4 Case study

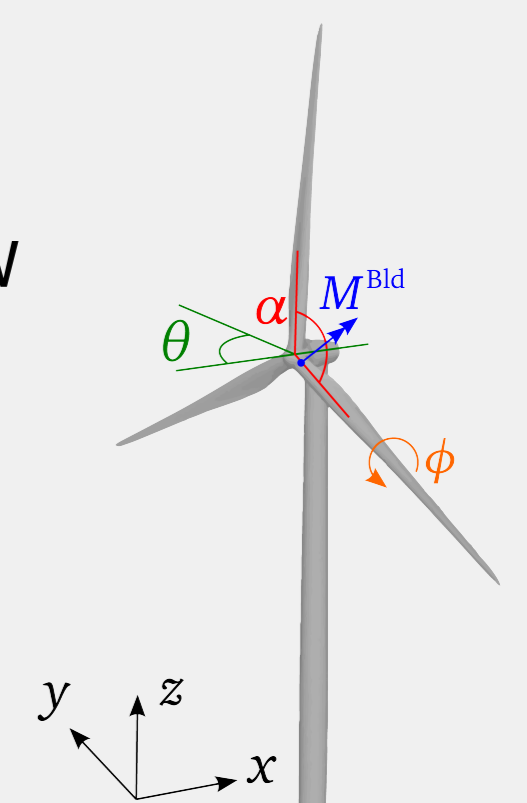
Complex **onshore wind turbine simulator** with control systems

- High-dimensional turbulent wind input: $\mathbf{v} : \mathcal{T} \rightarrow \mathbb{R}^{3 \times \nu_y \times \nu_z}$
- Quantity of interest: Flapwise blade root bending moment $M^{\text{Bld}} : \mathcal{T} \rightarrow \mathbb{R}$



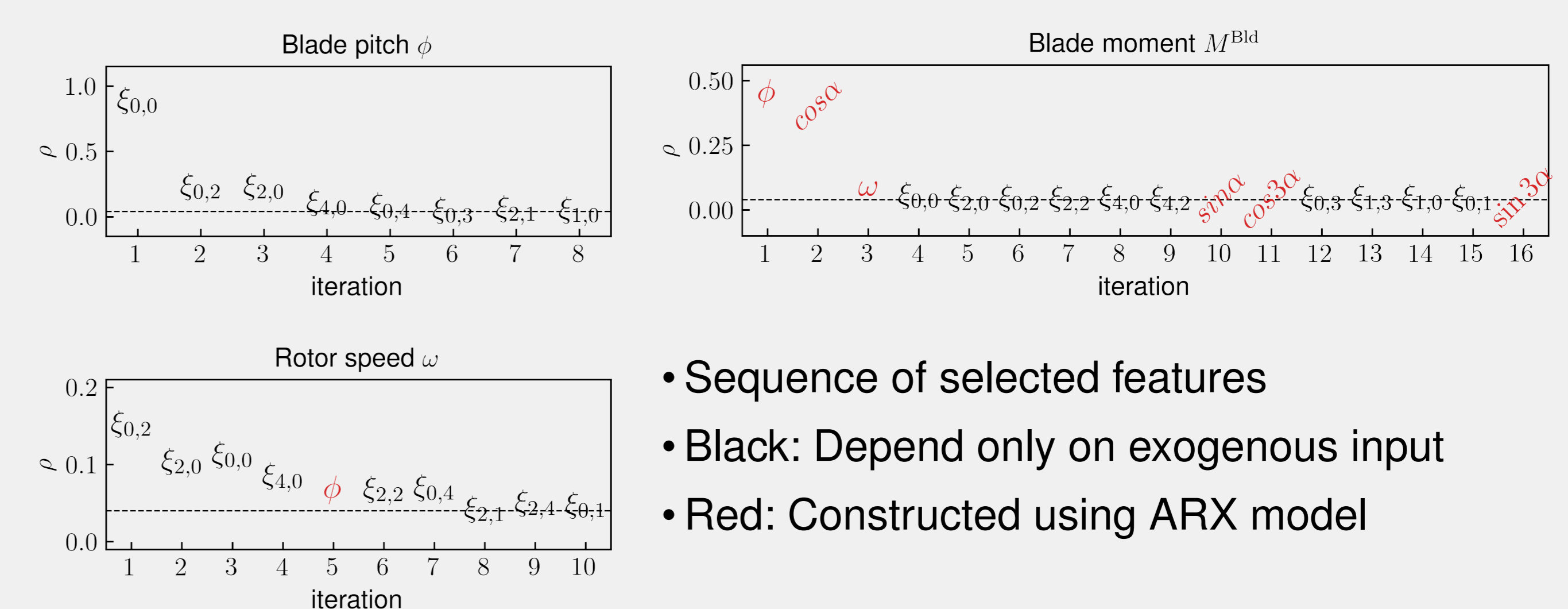
Computational model

Turbine	NREL 5-MW
Type	Onshore
Controller	ROSCO
Simulator	OpenFAST

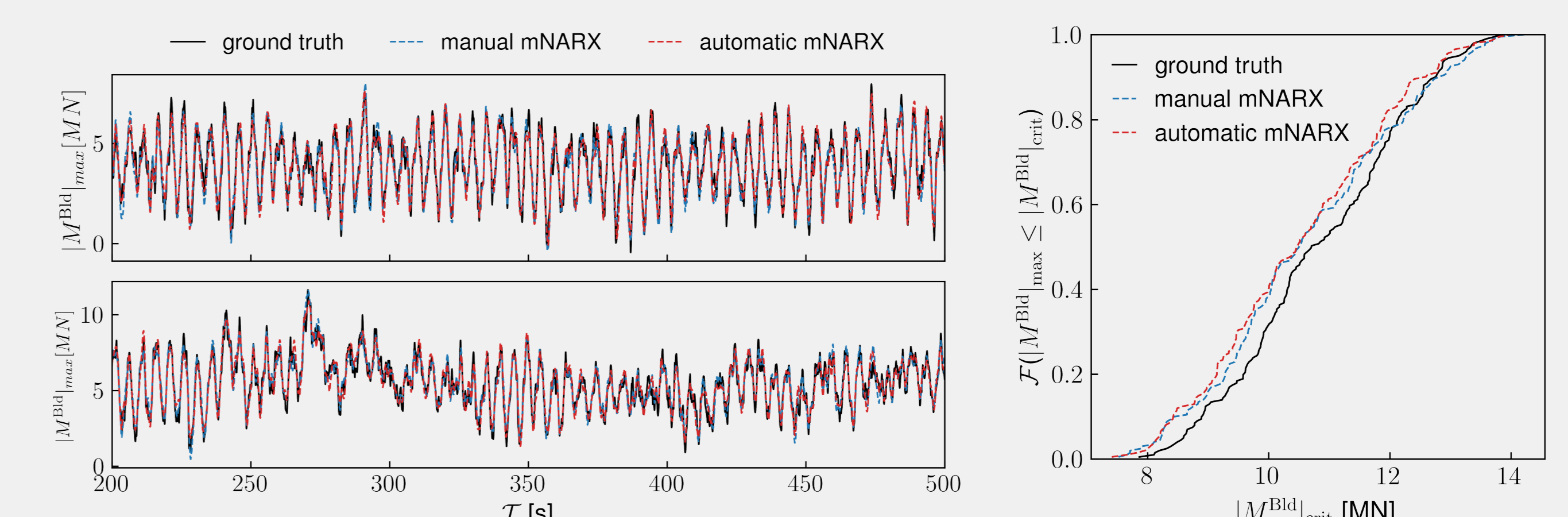


- Spatial compression of \mathbf{v} into spectral coefficients ξ

5 Results



- Sequence of selected features
- Black: Depend only on exogenous input
- Red: Constructed using ARX model



- (Left) Example traces: Simulated blade moment and the emulated ones using the manual and automatic mNARX surrogates
- (Right) CDF from 300 10-minute simulations and the corresponding emulated data using the manual and automatic mNARX surrogates

6 Discussion and Outlook

Discussion

- Multistep approach allows accurate emulation of complex dynamical systems
- Relevant features can be automatically selected using a measure of association
- Automatic selection reduces manual effort to construct an mNARX surrogate

Outlook

- Investigate the effect of the stopping criteria and measures of association
- Investigate the importance of the proxy surrogate to the final solution

References

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