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Neural Network Approach for Predicting Drum Pressure and Level in a Coal-fired Subcritical Power Plant

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10th European Conference on Coal Research and its Applications
Sept 15-17, 2014, University of Hull, UK

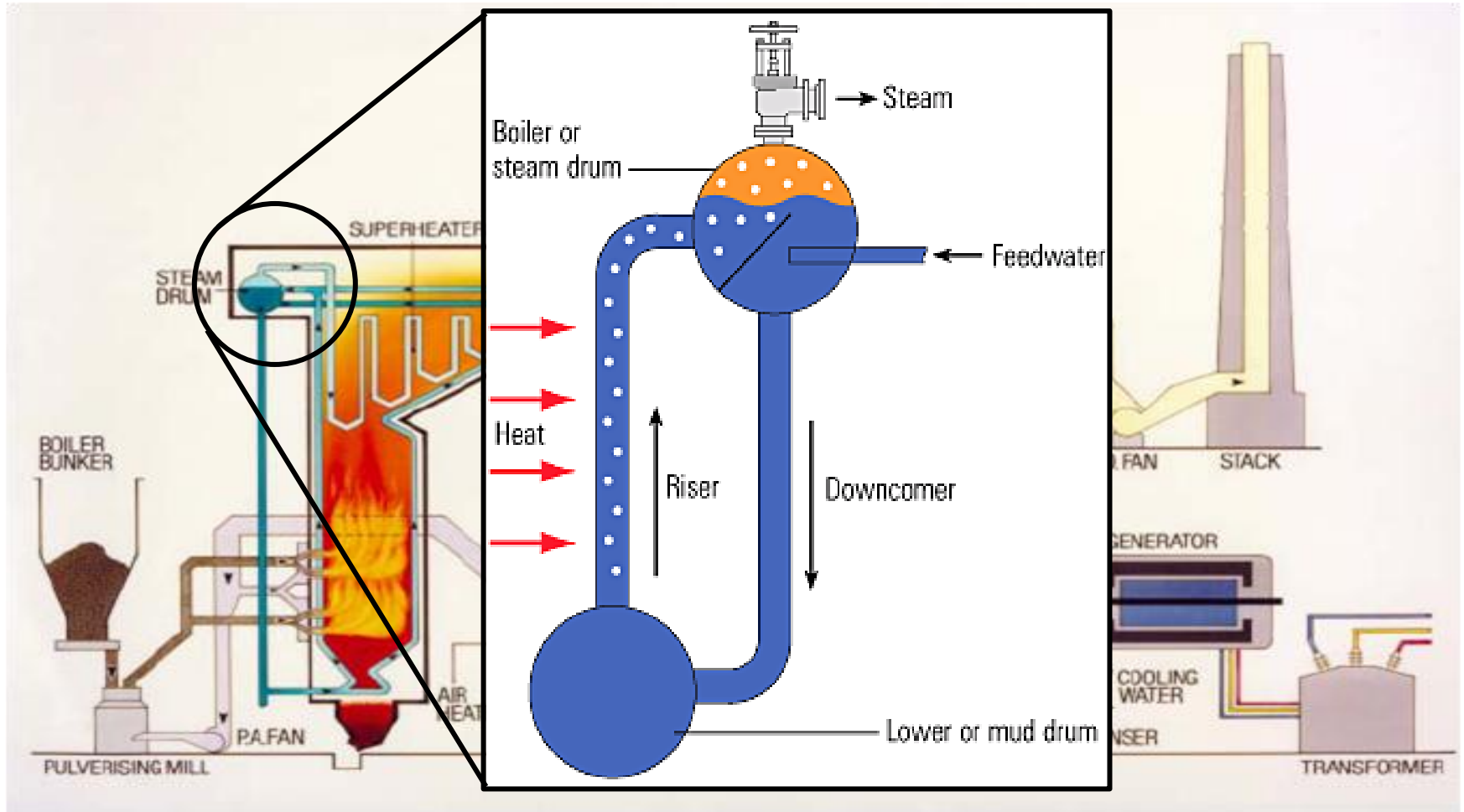
Outline

- Background
- Boiler Drum Data
- NARX Model
- NARX Model Training
- Results
- Conclusion and Future Work

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Background



Layout of Coal-fired Subcritical Power Plant¹

Background

- Boiler Drum is a critical component of a coal-fired subcritical power plant:
 - acts as energy reservoir in the plant enabling it to cope with short term load changes
 - dictates overall plant dynamics due to the slower dynamics compared to the steam turbine dynamics
- Due to tightening regulations, deserves study of the dynamics for operation and control purposes
- Data-based approach such as neural networks less laborious than physical modelling

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Boiler Drum Process

- Represented by a detailed first principle model according to Åström and Bell [3] using gPROMS tool.
 - Inputs: Feedwater flowrate, Steam flowrate and heat energy input
 - Outputs: Drum pressure and drum water level
- Thermodynamic properties of water/steam were obtained using IAPWS-95 formulation in Aspen Properties via COThermo interface.
- Thermodynamic property derivatives were obtained using polynomial approximations of steam table calculations from NIST REFPROP V9.1.

Boiler Drum Process

Global Conservation Equations

$$\frac{d}{dt}[\rho_s V_{st} + \rho_w V_{wt}] = m_{fw} - m_{st}$$

$$\frac{d}{dt}[\rho_s h_s V_{st} + \rho_w h_w V_{wt} - pV_t + M_t C_p T_{sat}] = Q + h_{fw} m_{fw} - h_s m_{st}$$

$$V_{st} + V_{wt} = V$$

Drum Dynamics

$$\frac{d}{dt} \rho_s V_{sd} = \alpha_r m_r - m_{sd} - m_{cd}$$

$$m_{sd} = \frac{\rho_s}{T_d} (V_{sd} - V_{sd}^0) + \alpha_r m_{dc} + \alpha_r \beta (m_{dc} - m_r)$$

$$V_{wd} = V_{wt} - V_{dc} - (1 - \bar{\alpha}_v) V_r$$

$$m_{cd} = m_{fw} \frac{h_w - h_{fw}}{h_s - h_w} + \frac{1}{h_s - h_w} [\rho_s V_{sd} \frac{dh_s}{dt} + \rho_w V_{wd} \frac{dh_w}{dt} - (V_{sd} + V_{wd}) \frac{dp}{dt} + m_d C_p \frac{dT_{sat}}{dt}]$$

$$L_d = \frac{V_{wd} + V_{sd}}{A_d}$$

Downcomer-Riser Dynamics

$$\frac{d}{dt} [\rho_s \bar{\alpha}_v V_r + \rho_w (1 - \bar{\alpha}_v) V_r] = m_{dc} - m_r$$

$$\begin{aligned} \frac{d}{dt} [\rho_s h_s \bar{\alpha}_v V_r + \rho_w h_w (1 - \bar{\alpha}_v) V_r - pV_r + M_r C_p T_{sat}] \\ = Q + m_{dc} h_w - (\alpha_r (h_s - h_w) + h_w) m_r \end{aligned}$$

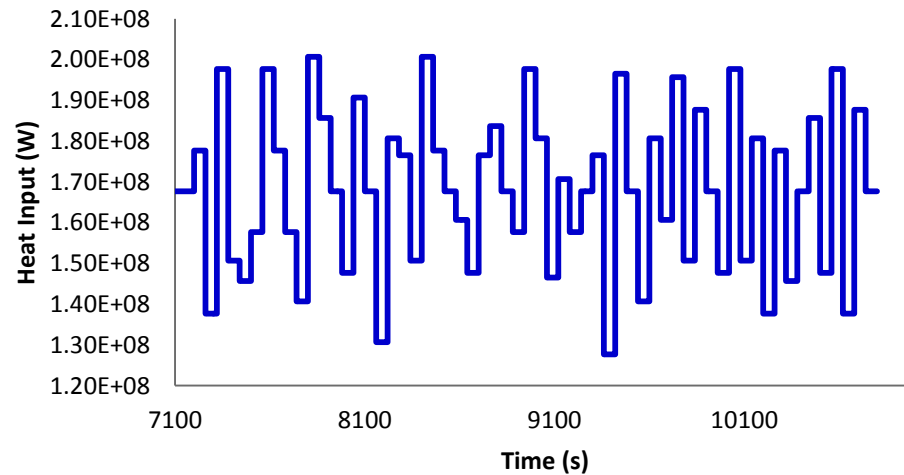
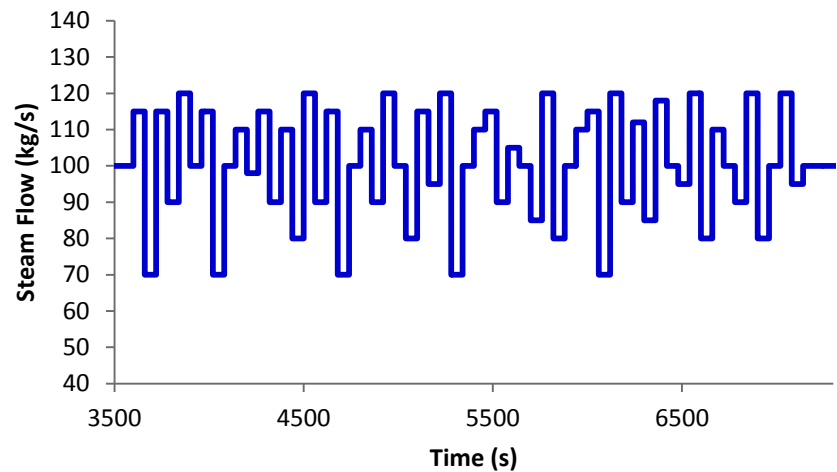
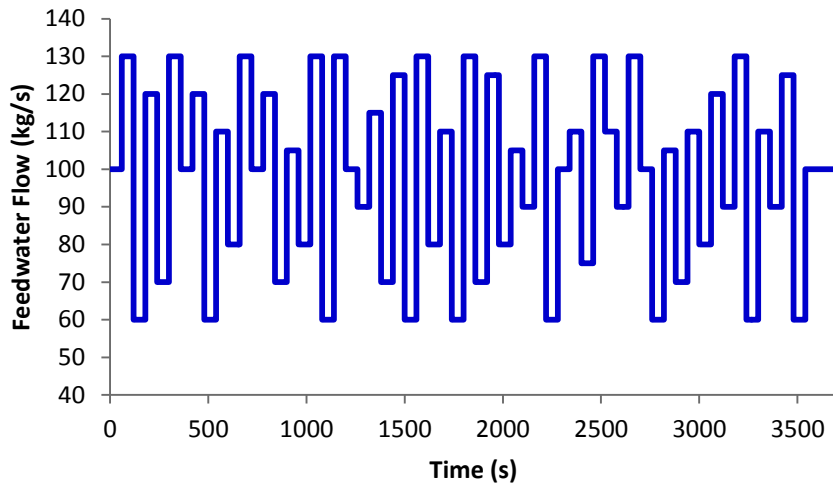
$$\bar{\alpha}_v = \frac{\rho_w}{\rho_w - \rho_s} \left[1 - \frac{\rho_s}{\alpha_r (\rho_w - \rho_s)} \ln \left(1 + \alpha_r \frac{\rho_w - \rho_s}{\rho_s} \right) \right]$$

$$m_{dc} = \sqrt{\frac{2g\rho_w \bar{\alpha}_v A_{dc} V_r (\rho_w - \rho_s)}{k}}$$

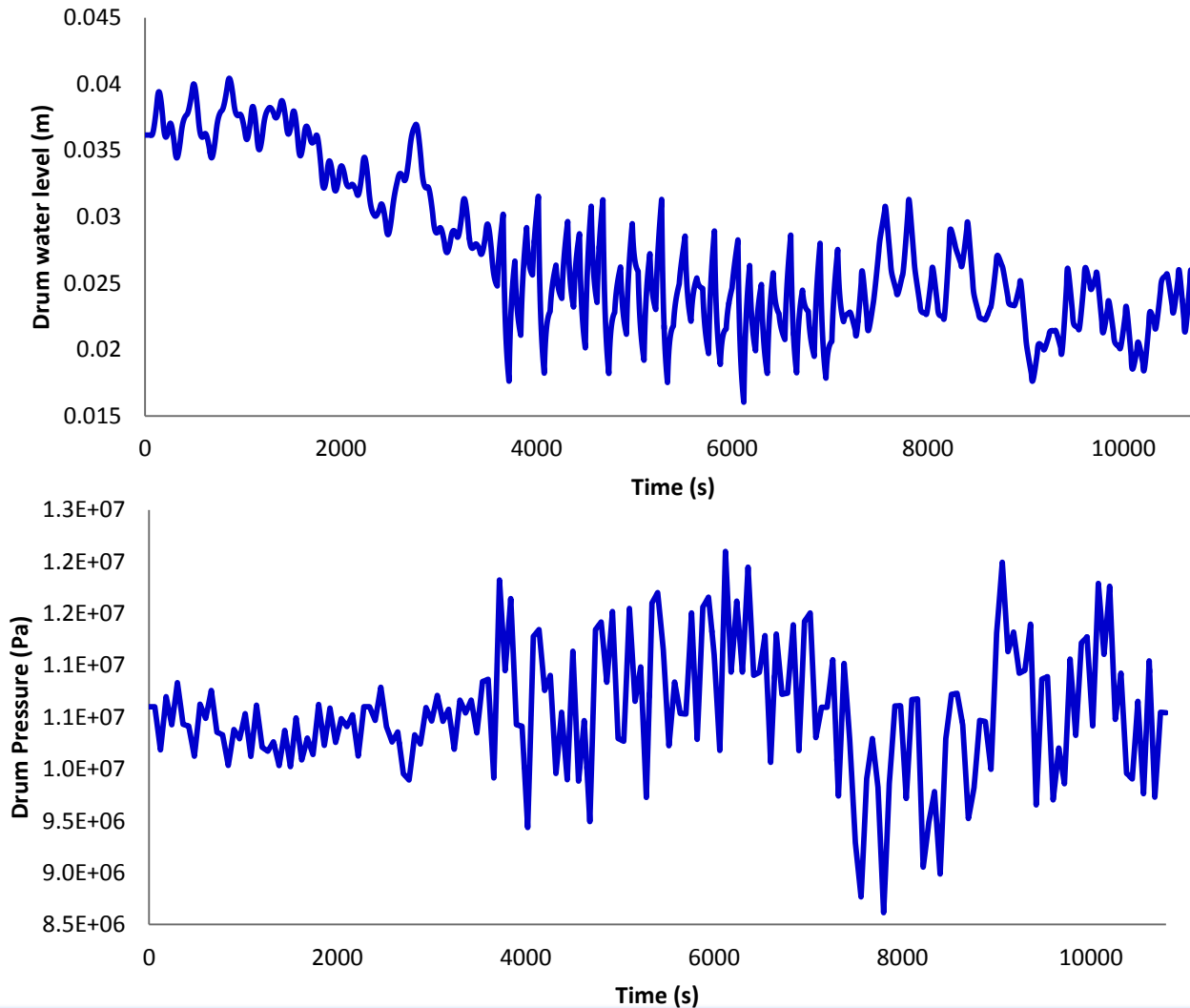
Boiler Drum Data

- Data obtained at open loop conditions using the developed model
- The system is excited by perturbing the inputs in succession with a series of step changes in no particular order
- Perturbation in each input is sustained for an hour resulting to a total test period of 3 hours (10800 seconds).
- Other inputs are maintained at their equilibrium value while perturbing the other.

Boiler Drum Data



Boiler Drum Data

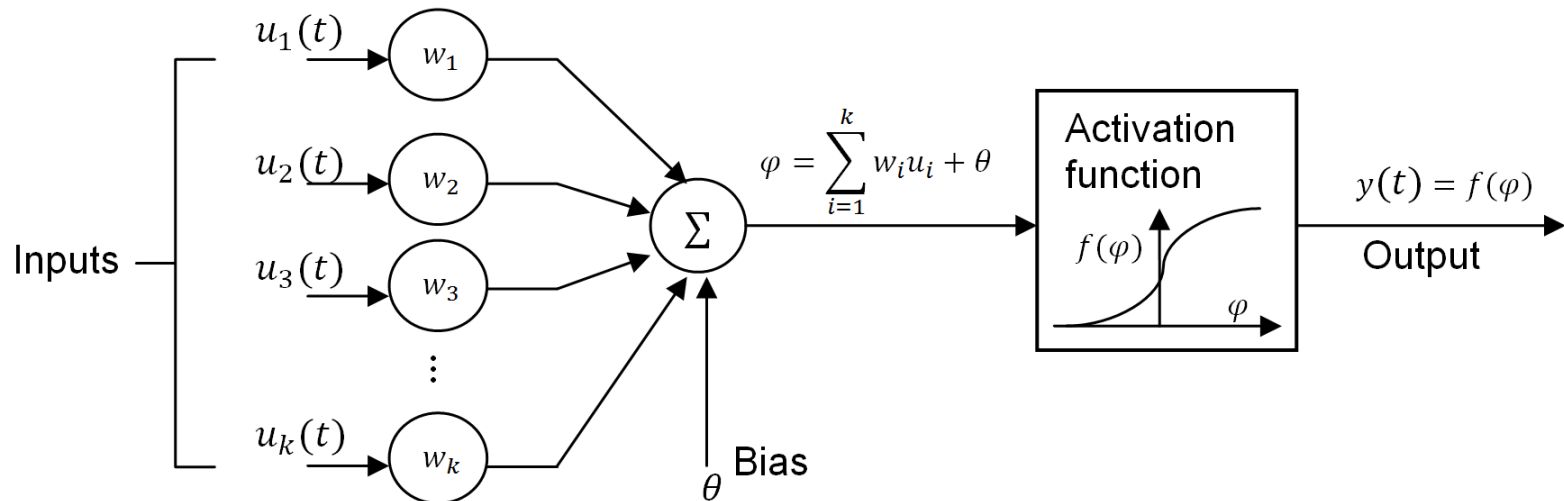


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Neural Networks (NNs)

- Computational paradigm inspired from the structure of biological neural networks and their way of encoding and solving problems.
- Able to identify underlying highly complex relationships based on input-output data only.



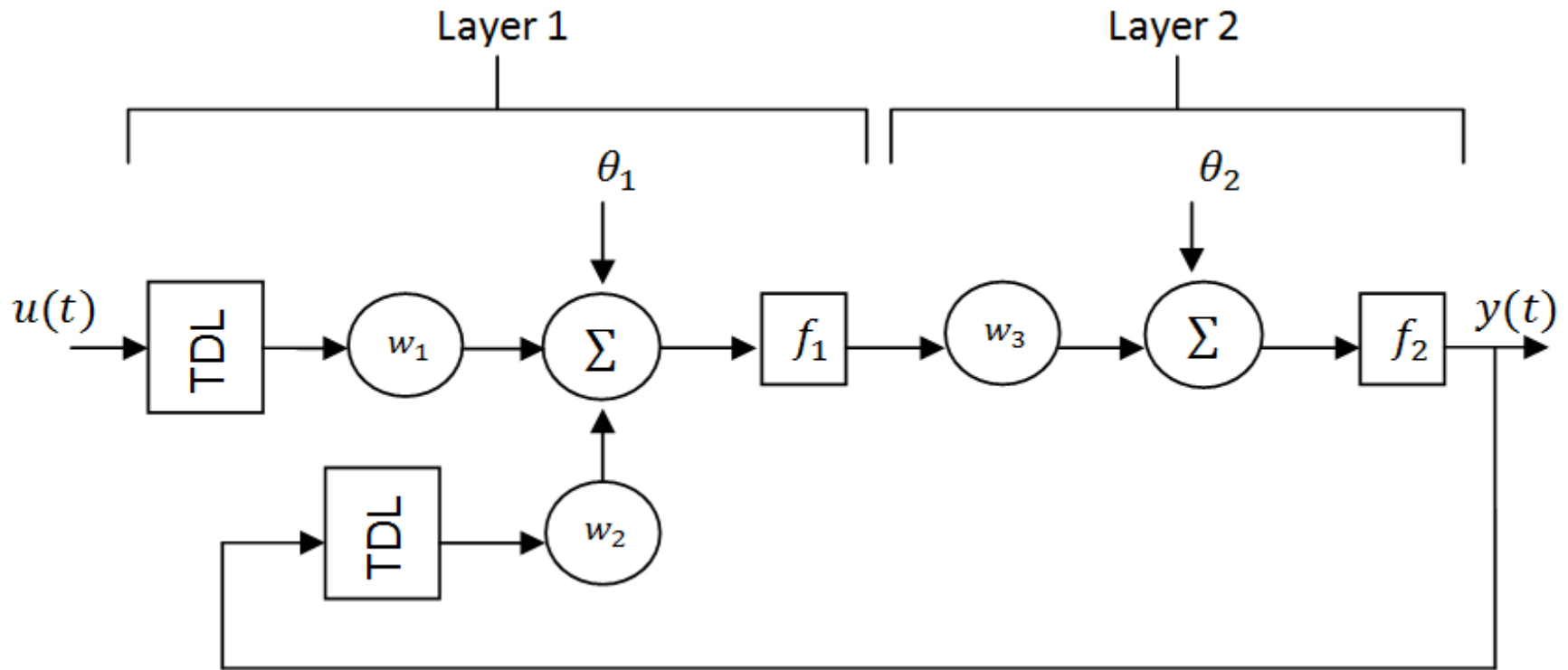
Neural Networks (NNs)

- NN model capable of reproducing time-dependent (dynamic) data takes into account the time variable by means of a memory process.
- This is done using NN architectures with feedback connections among neurones and time delay lines (TDL)
- NARX NN is a typical NN with feedback connections enclosing several layers of the network and a TDL.

NARX Model

➤ NARX Model is defined by the Equation:

$$y(t) = f[y(t-1), y(t-2), \dots, y(t-n_y), u(t-1), u(t-2), \dots, u(t-n_u)]$$

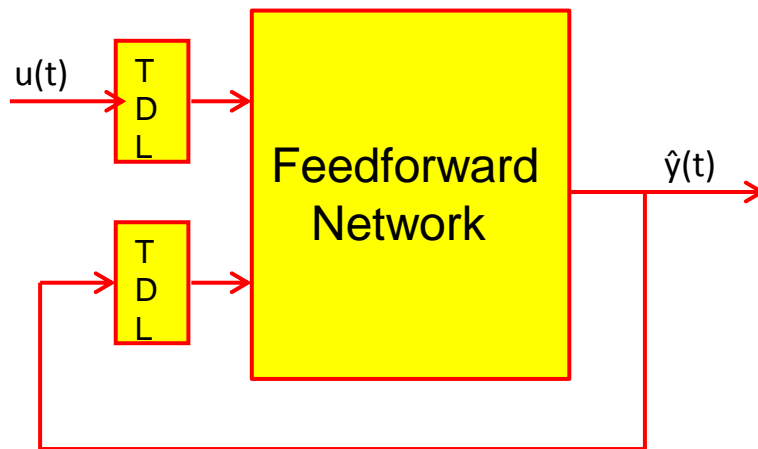


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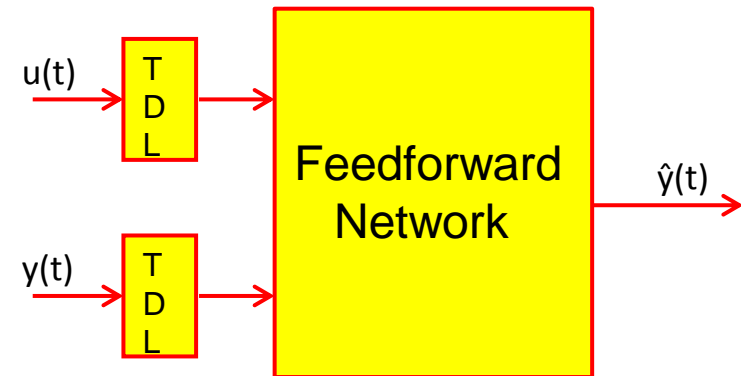
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NARX Model Training

- Training complicated due to the feedback loop; ideally dynamic training algorithm which is complex is needed



(a) Parallel Architecture
(Standard NARX Architecture)



(b) Series-Parallel Architecture

- With (b), less complex static training algorithm is used for training.

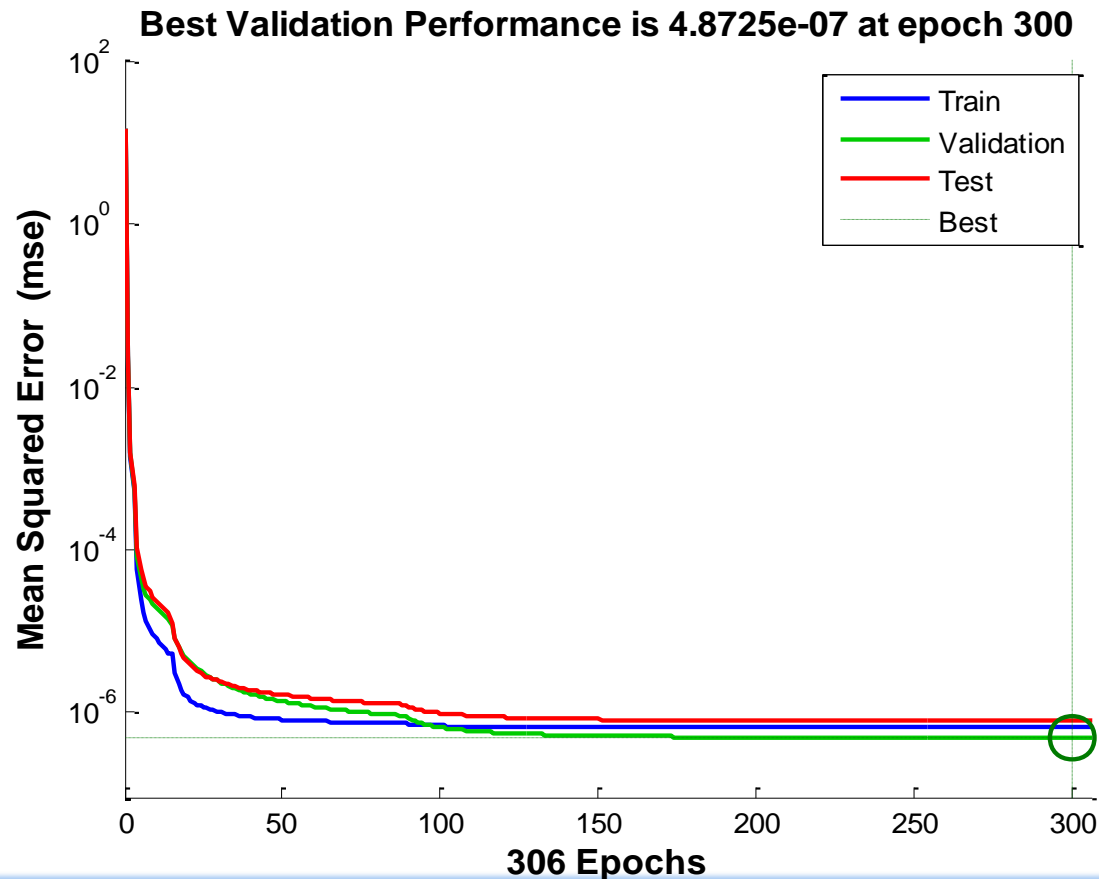
NARX Model Training



- Training with simulated data from the detailed physical model.
- Early stopping technique used to avoid overfitting: Data divided into training (70%), validation (15%) and testing (15%) data.
- Levenberg-Marquardt training algorithm used (*trainlm*) with mean squared error (MSE) performance function.
- 100 neurones in the hidden layer each with a sigmoid transfer function

NARX Model Training

$$\text{MSE} = \frac{1}{N} \sum_{i=0}^N (z_i - y_i)^2 \rightarrow \min$$

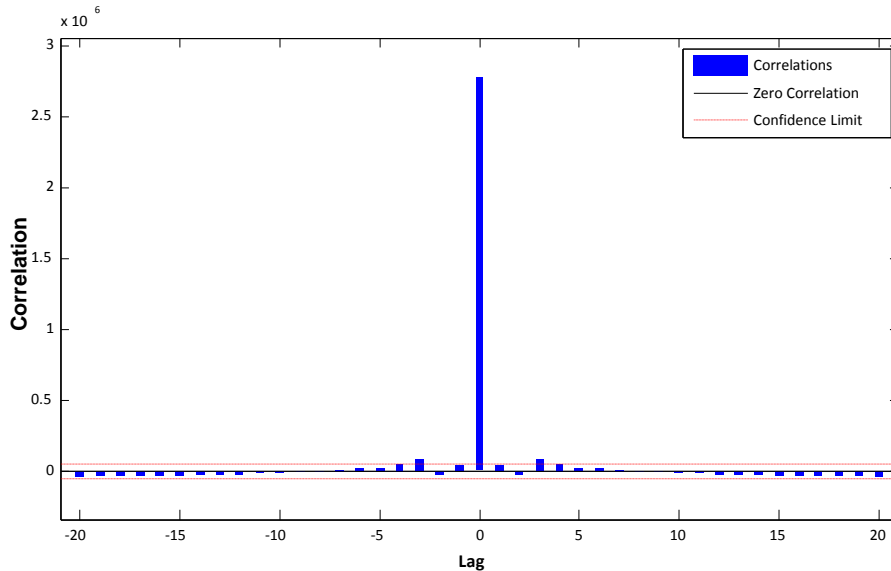


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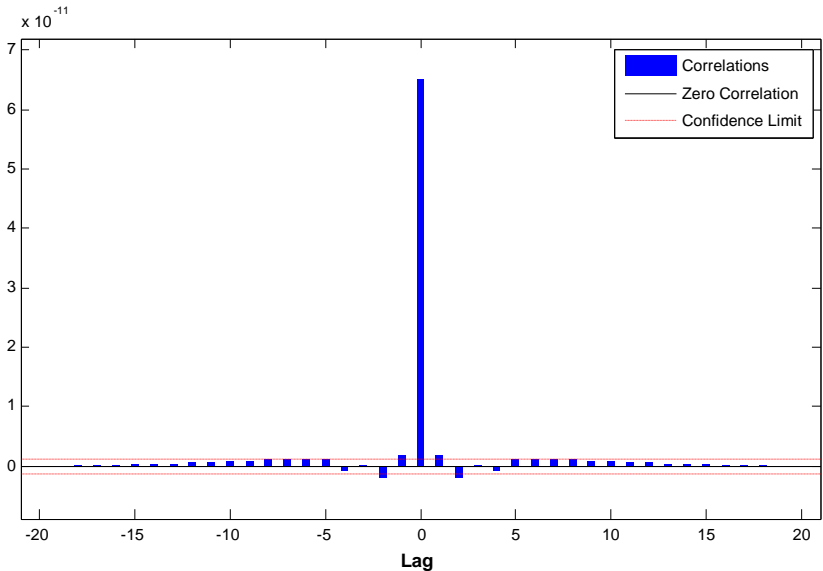
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Results

Error Autocorrelation Plot



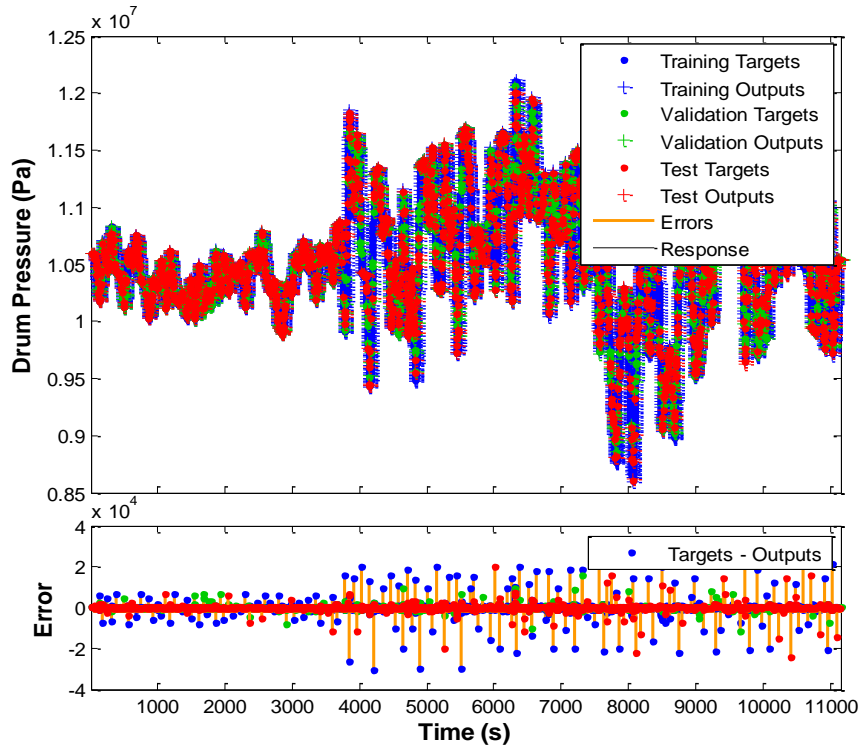
Drum Pressure Prediction Error
Autocorrelation Plot



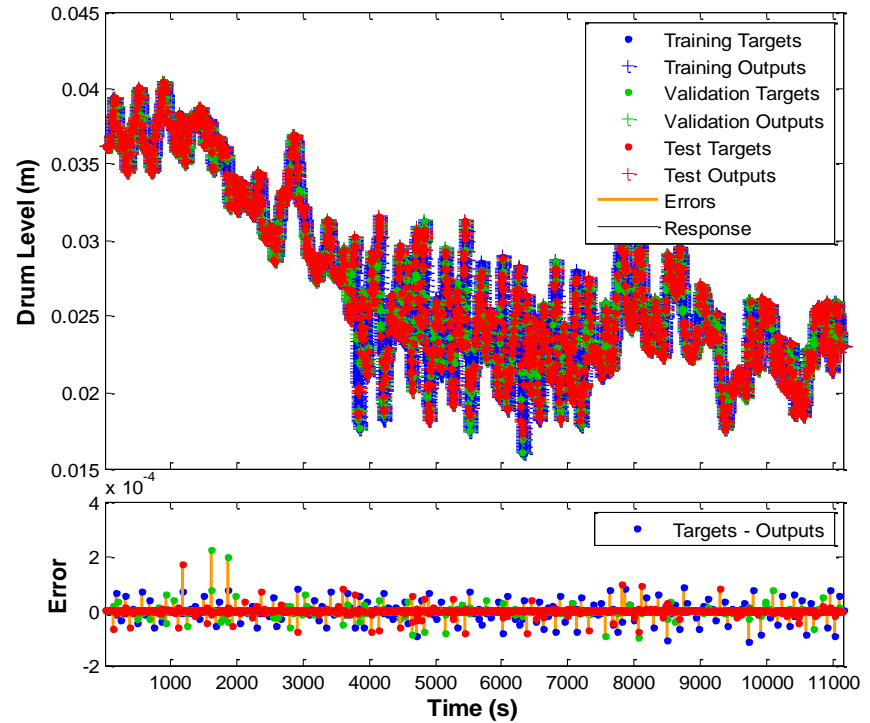
Drum Level Prediction Error
Autocorrelation Plot

- The plots show reliable estimate of the network parameters, weights and biases

Results



Output/Target Comparison (Drum Pressure)

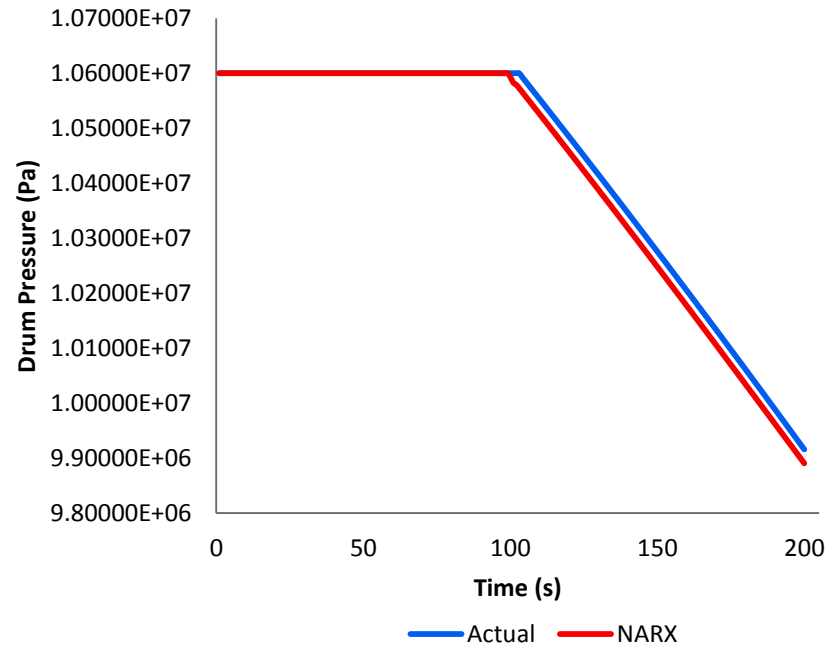
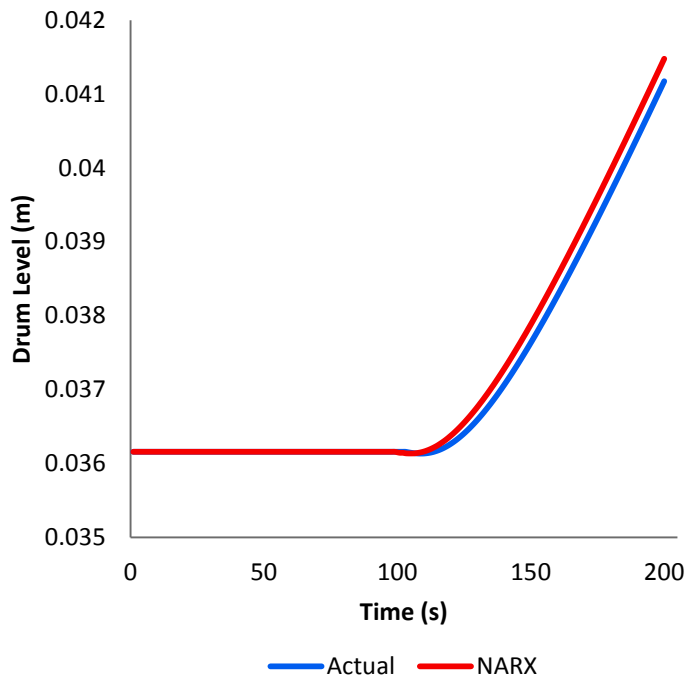


Output/Target Comparison (Drum Level)

Results

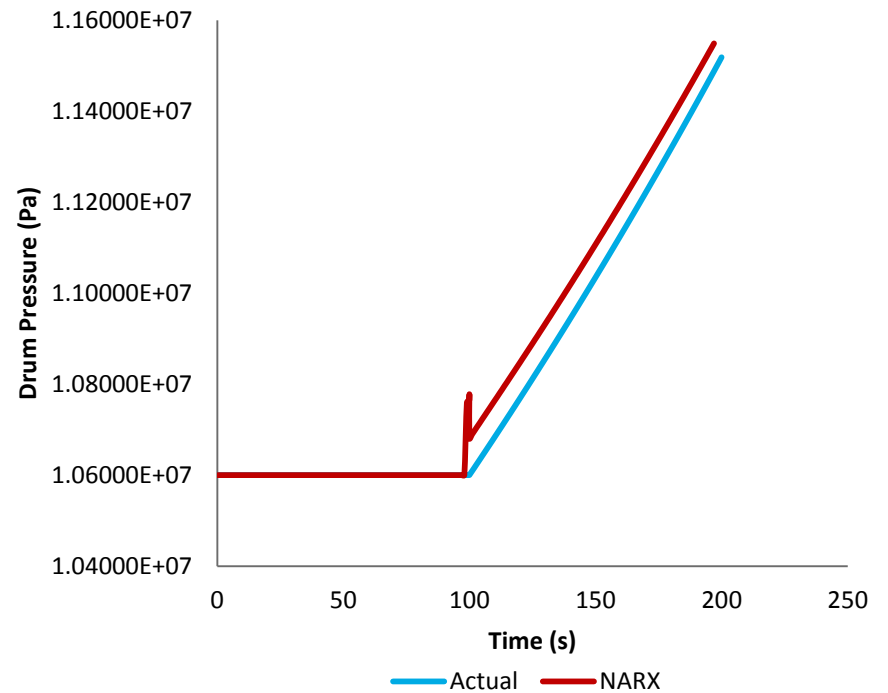
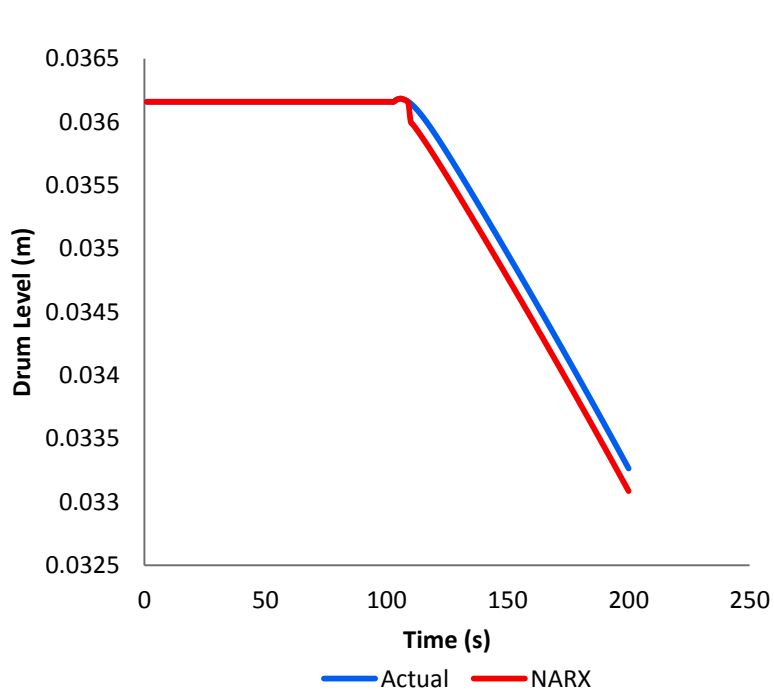
Step Change Test: Performed on one input at a time while other inputs remain unchanged

➤ +30 kg/s step change in feedwater flowrate



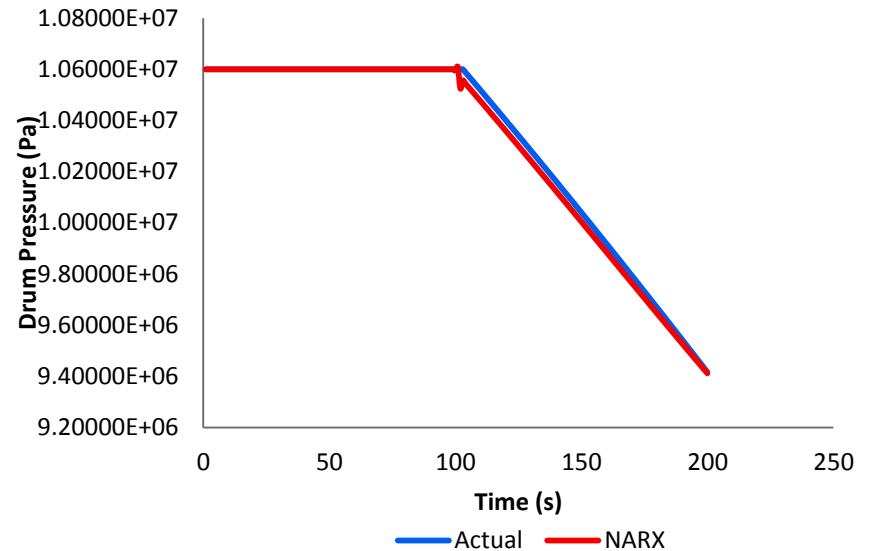
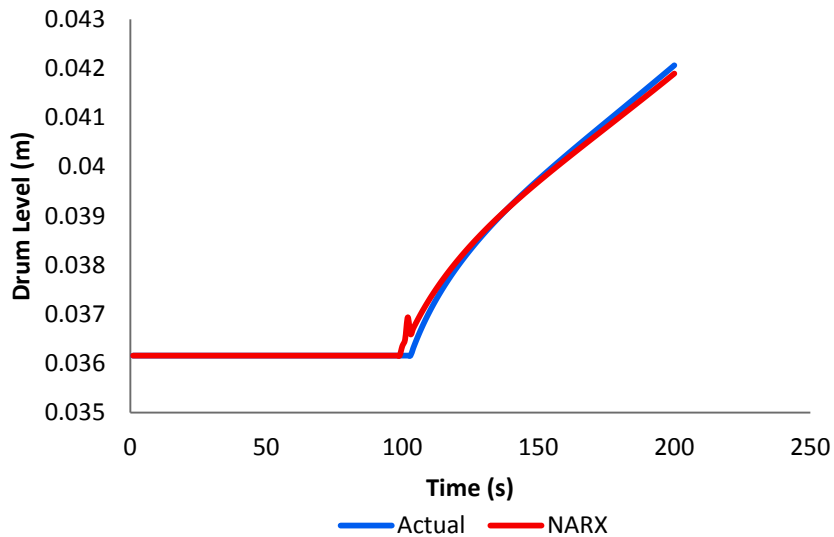
Results

➤ +10 MW step change in Heat Input



Results

➤ +10 kg/s step change in Steam Flowrate



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Conclusion and Future Work

- NARX NN can be used to obtain reliable dynamic model of a boiler drum from the plant input-output operating data only.
- Use of NARX NN avoids the rigours of reliable parameter estimation often needed in modelling from first principle.
- As it is the case with all empirical models, NARX NN models are only reliable when they are used within the conditions that they were trained.
- The NARX NN model developed in this study is subject to the inherent deficiencies in Åström and Bell [3] model which was used to obtain the training data.
- Development of model predictive control (MPC) of the boiler drum based on the NARX NN model is on-going.

References

- 1) Illustrations and text are taken from the Spirax Sarco website 'Steam Engineering Tutorials' at <http://www.spiraxsarco.com/resources/steam-engineering-tutorials.asp>. Illustrations and text are copyright, remains the intellectual property of Spirax Sarco, and have been used with their full permission
- 2) Åström, K.J. and Bell, R.D. Drum-boiler dynamics. Automatica 2000; 36: 363-378.

Acknowledgement

Natural Environment Research Council (NERC), UK



EU FP7 Marie Curie



Multiphase Flow Measurement Research Group,
South East University, Nanjing, China



Questions???

Thank you for your Attention!

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