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XXXIV Cycle - II year presentation

Hysteresis Modeling in Iron-Dominated Magnets based on a Deep Neural Network Approach

RESEARCH ACTIVITY CONTEXT

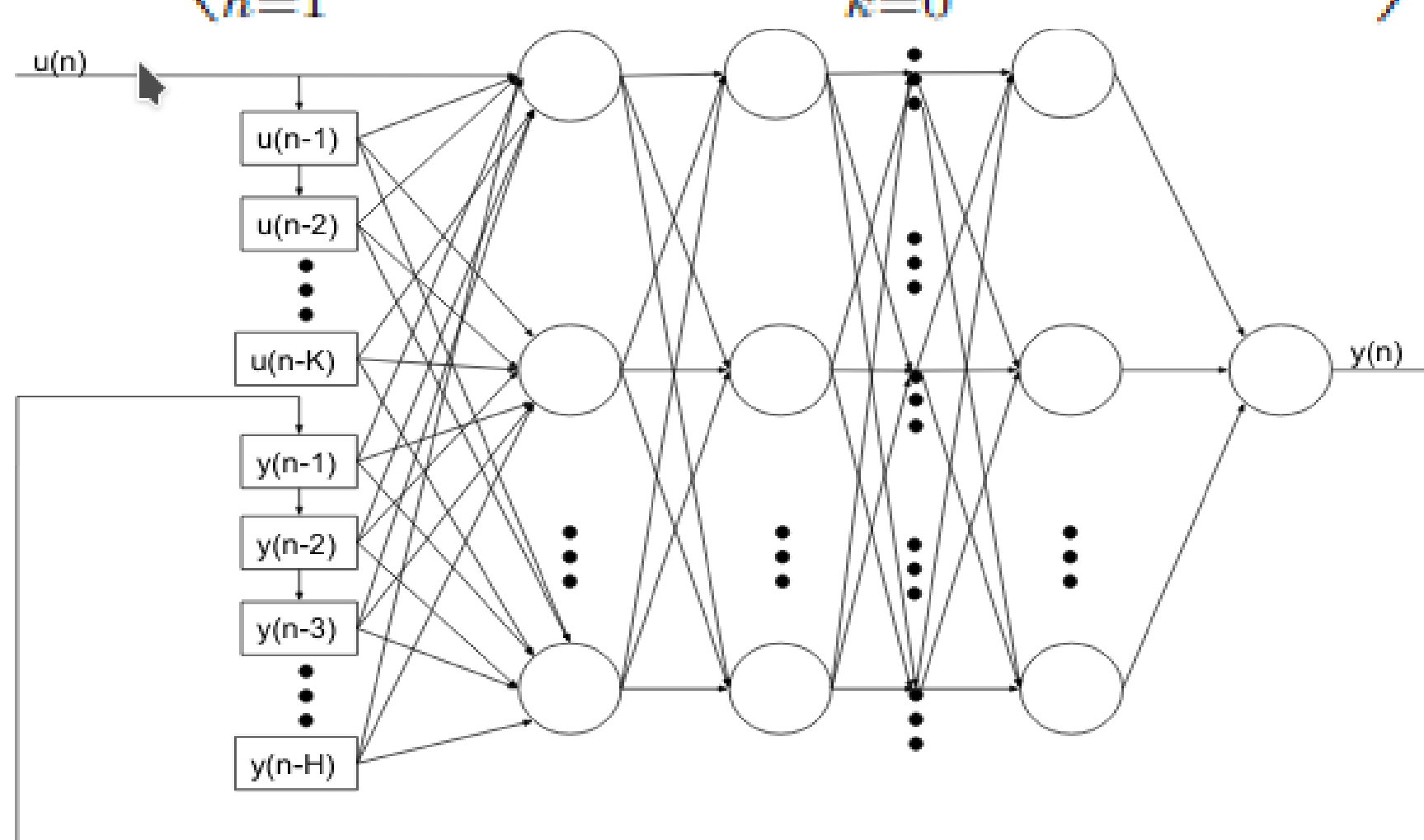
Modeling of quasi-static and dynamic hysteresis loops is one of the most challenging topics in computational magnetism, mainly due to the strong non-linearity and history dependency shown by ferromagnetic materials. The complex excitation current waveforms $I(t)$, used in particle accelerators (PAs) are still an open focus of scientific interest because $B(I)$ becomes much more complex and hard to predict. In PAs, the beam is accelerated by radio frequency cavities which generate a bending field, increasing in proportion to the beam momentum. Accurate knowledge of the magnetic field $B(t)$ at any given time during a magnetic cycle is critical for beam control, power supply control, beam diagnostics, and qualitative feedback to operators. The required accuracy is typically 0.01%.

I worked on tuning a deep neural network to fit directly the magnet response, by avoiding complementary physical models.

NEURAL NETWORK ARCHITECTURE

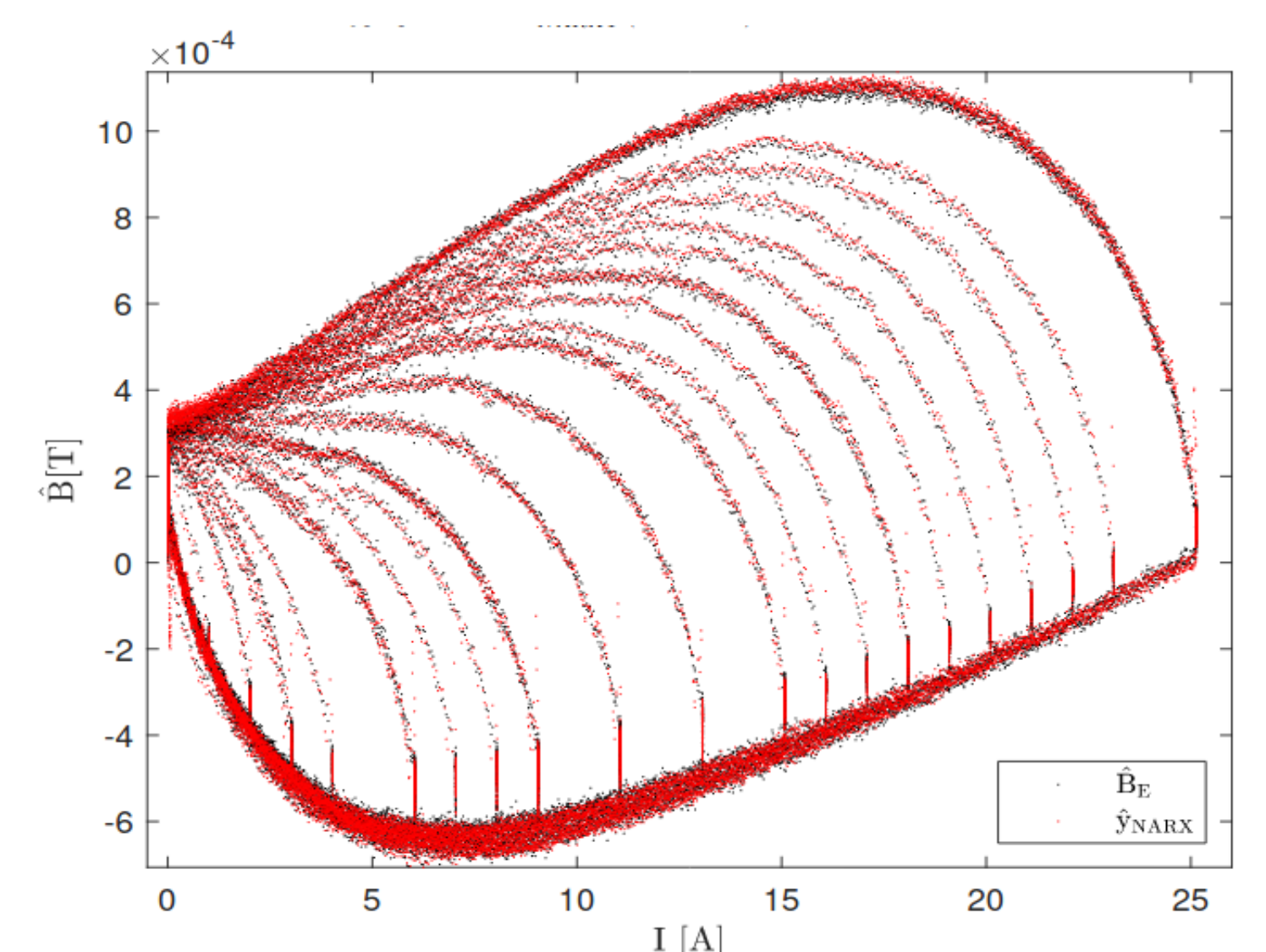
Different architectures are considered and selected according to a compromise between the accuracy of the field estimation and the level of complexity of the network. The best performing architecture results a *Nonlinear Autoregressive Exogenous Neural Network* (NARX), which relies on temporal feedback to capture the underlying physics. The results of tests carried out on a dedicated experimental setup, taken as a reference

$$a_j^1(n) = \phi^{A_1} \left(\sum_{h=1}^H W_{jh}^O y(n-h) + \sum_{k=0}^K W_{jk}^I u(n-k) \right)$$



RESULTS

The NARX networks achieve a bestcase NRMSE of 0.006%. Thanks to having memory of past outputs, the NARX networks are shown to be able to reconstruct very accurately the dynamic



$$b) \quad RMSE(y, D^{test}) = \sqrt{\frac{\sum_{n \in N} (y_i(n) - B^{test}(n))^2}{|N|}} \quad NRMSE(y, D_E) = \frac{RMSE(y, D_E)}{B_{max}} \cdot 100$$

CONCLUSIONS

We tested our network on the raw datasets. We found that NARX networks achieve in general the required level of performance i.e. an NRMSE better than 0.01%. Such excellent performance paves a very promising way for future applications in this context.

COLLABORATIONS

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FUTURE WORK

The plans for the future are to train and test NARX networks on a wider variety of excitation waveforms, such as e.g. sequences of cycles with flat-tops increasing or decreasing randomly, which are representative of the most challenging actual operating conditions of accelerator magnets. As part of the renovation of the real-time magnetic measurement systems currently ongoing at CERN, we are implementing in FPGA hardware a real-time version of the NARX networks that will be able to carry out a continuous field prediction, in parallel to the measurements. **For contacts:** vincenzo.dicapua@unina.it vincenzo.di.capua@cern.ch