
Master 2 internship proposal (2023 – 2024)

Machine Learning-Based Material Modeling for Electric Motor Design

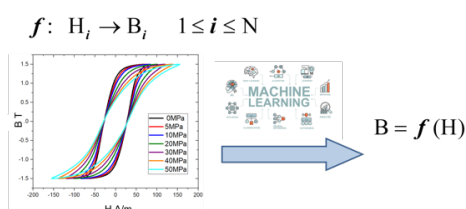
Start : Available starting from February 1, 2024
Duration : 4 or 6 months
Location : L2EP laboratory (Laboratory of electrical engineering and power electronics), University of Lille
Salary : Around 600€
Supervisors : Zuqi Tang (zuqi.tang@univ-lille.fr), Abdelkader Benabou (abdelkader.benabou@univ-lille.fr)
Application : CV and cover letter

Context

Efficiently optimizing the energy performance of modern electrical motors, particularly in industrial and electric mobility applications, relies on the availability of precise design tools. A critical component in achieving this goal is accurate material modeling, specifically for electrical steels used in the magnetic circuits of these motors. These materials' properties are instrumental in defining overall performance and energy efficiency during the energy conversion. However, the intricacies of manufacturing processes and the demanding operating conditions of these electrical motors require the consideration of the multi-physical behaviors exhibited by electrical steels.

In contemporary applications, electrical steels' mechanical and thermal stresses significantly impact properties crucial to energy conversion, such as magnetic constitutive laws and iron losses. The complex multi-physical couplings inherent to electrical steels, including magneto-mechanical and magneto-thermal interactions, often lead to property degradation, ultimately affecting motor performance. Conventional approaches to modeling these materials under manufacturing processes and operating conditions tend to result in complex and computationally intensive models. These models could be more practical for design tools, especially during optimization procedures.

Objective



This internship aims to address these challenges by focusing on innovative approaches to material modeling, specifically through machine learning techniques. The primary objective is to compare and evaluate different conventional machine-learning techniques for constructing surrogate models of electrical steel properties. Two key scenarios will be explored: one considering the influence of mechanical effects and the other excluding them.

Work steps

- 1 Evaluating and comparing the performance of various machine learning methods for constructing surrogate models of electrical steel properties.
- 2 Investigating the impact of mechanical effects on the accuracy of material modeling.
- 3 Identifying the most suitable machine learning approach for creating surrogate models that strike a balance between accuracy and computational efficiency.

Keyword:

Machine learning techniques, material modeling, multi-physics

References

- [1] H. H. Saliyah, D. A. Lowther and B. Forghani, "Modeling magnetic materials using artificial neural networks," in *IEEE Transactions on Magnetics*, vol. 34, no. 5, pp. 3056-3059, Sept. 1998, doi: 10.1109/20.717715.
- [2] R. Gong, A. Benabou, and Z. Tang, "Computer-Aided Measurement method of Hysteresis Loop based on Convolution Neural Network," *SGE SGE 2021, Nantes, France, 07/2021*.
- [3] A. Khan, S. Ceesay, Y. -P. Teng, R. Wang, H. Yue and D. Lowther, "Generalizable Deep Neural Network Based Multi-Material Hysteresis Modeling," *CEFC 2022, Denver, CO, USA, 2022*, pp. 1-4.