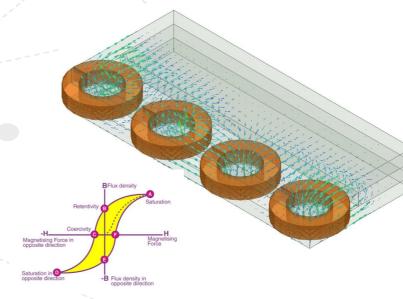
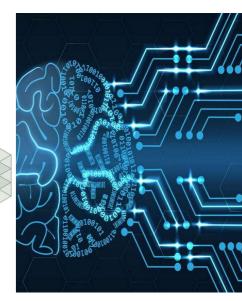


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Leistungselektronik und elektrische Antriebstechnik

Nachwuchsforschungsgruppe





# MAGNET CHALLENGE 2023 EXPLORATORY DATA ANALYSIS (EDA)

**Project Kick-Off** 

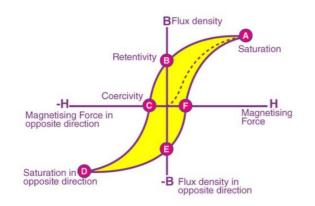
4/24/23

Wilhelm Kirchgässner

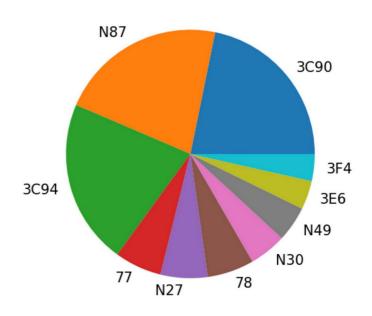


#### Data set in a nutshell

- 186747 operation points / samples
  - 2.8 GB disk space (serialized and compressed)
- Each sample consists of
  - Frequency, Temperature, B-curve (inputs)
  - Power loss, H-curve (targets)
  - A class label (material)
- Each curve consists of 1024 points

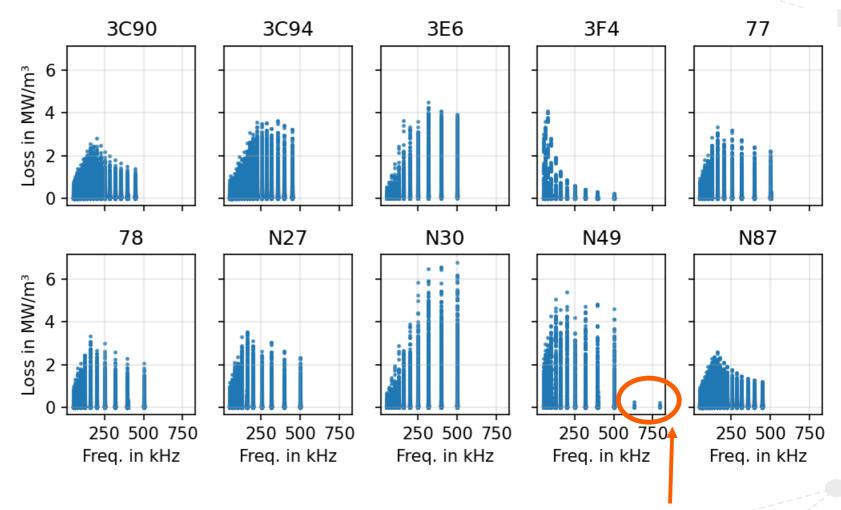


#### Size portions per material



#### **Distributions**

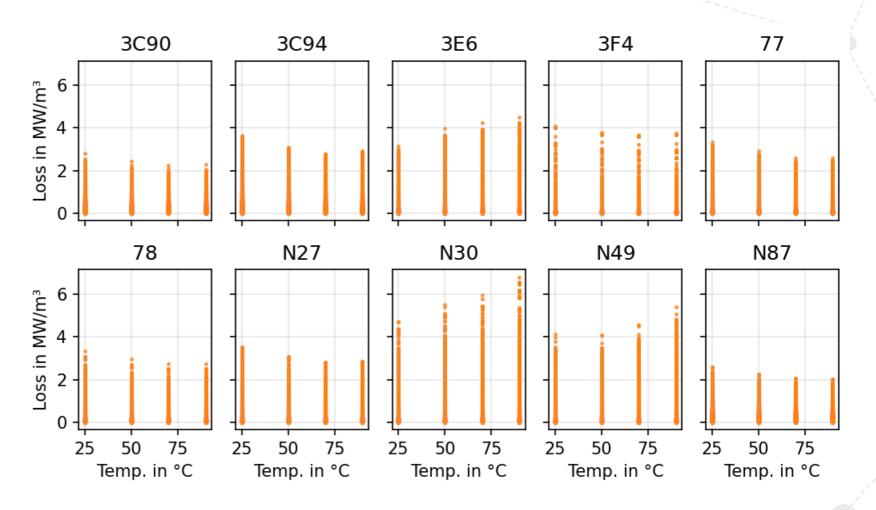
- Linear increase of max losses
- 1/freq bound for higher frequencies



Highest frequencies only for one material

### **Distributions**

Only four distinct temperatures



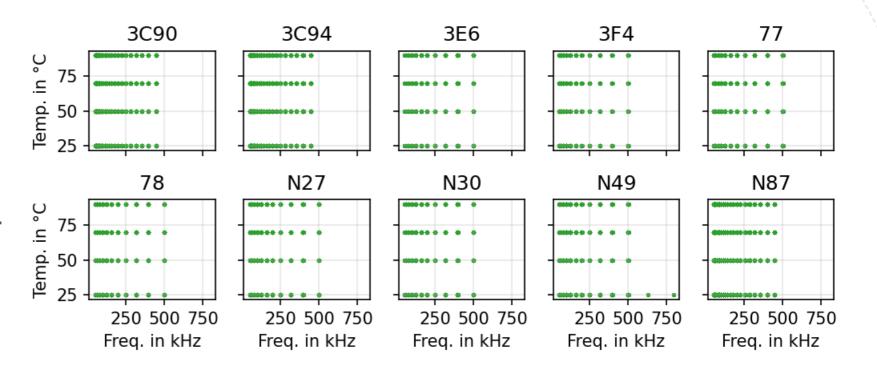


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#### **Distributions**

- Highest frequencies only for coldest temp.
- Materials with less portion size have fewer different freqs. at the top range

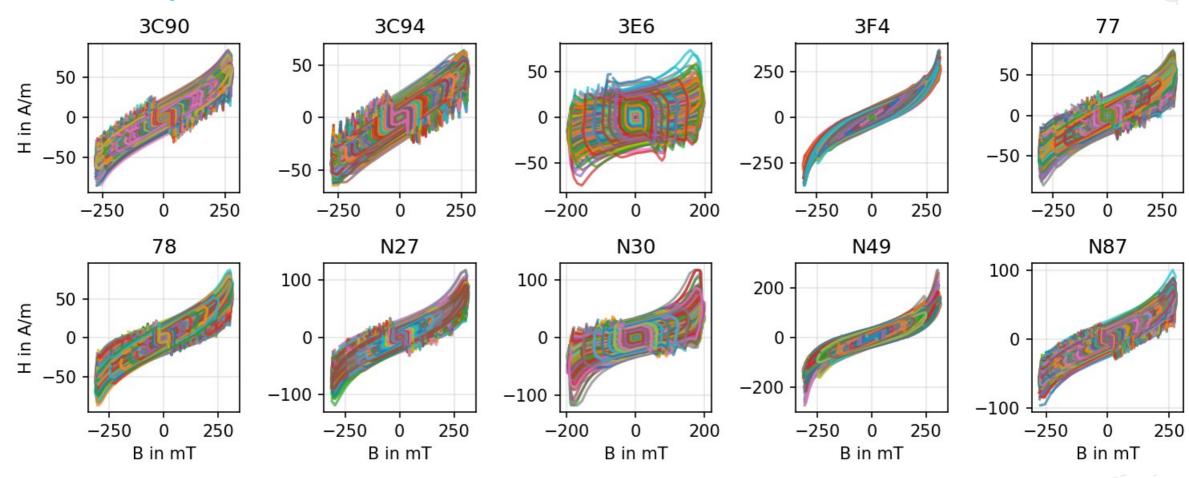




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# **B-H-Curves per material**

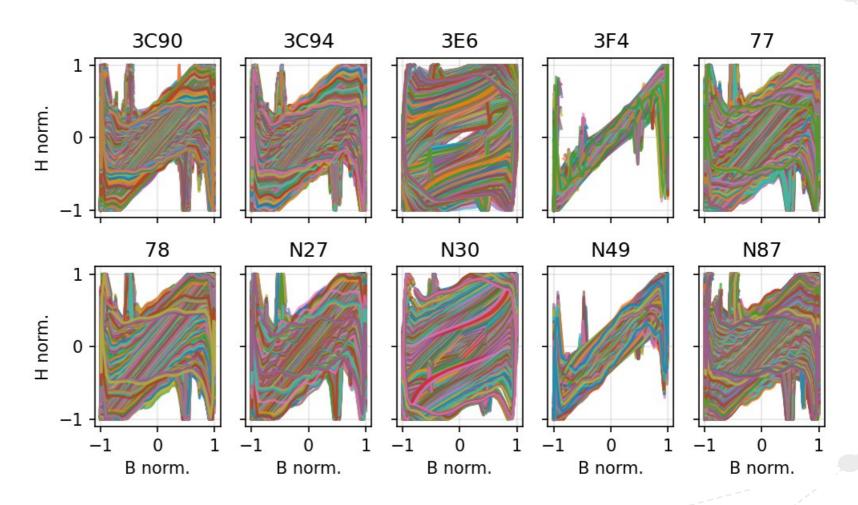
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# **B-H-Curves per material - Normalized**

- Materials with lower max(B) have less distinct fins
- Materials with high max(H) have the most concise fins

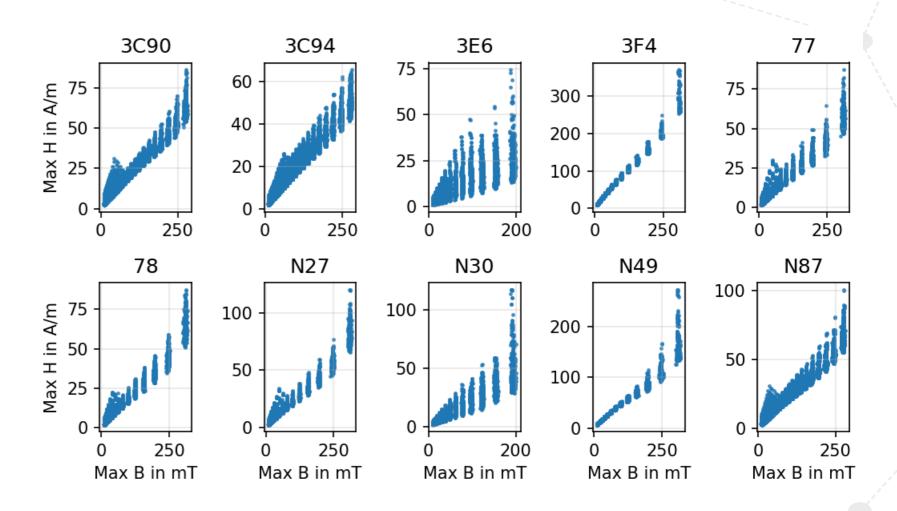


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# **Distributions (2)**

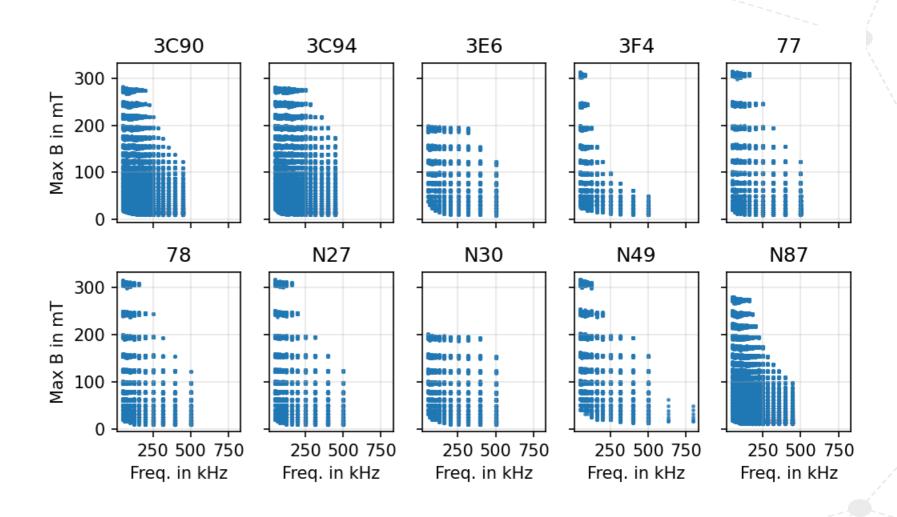
 Somewhat linear (except 3E6)



# **Distributions (2)**

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- Hyperbolic decrease of max(B) over frequency
- Different for each material



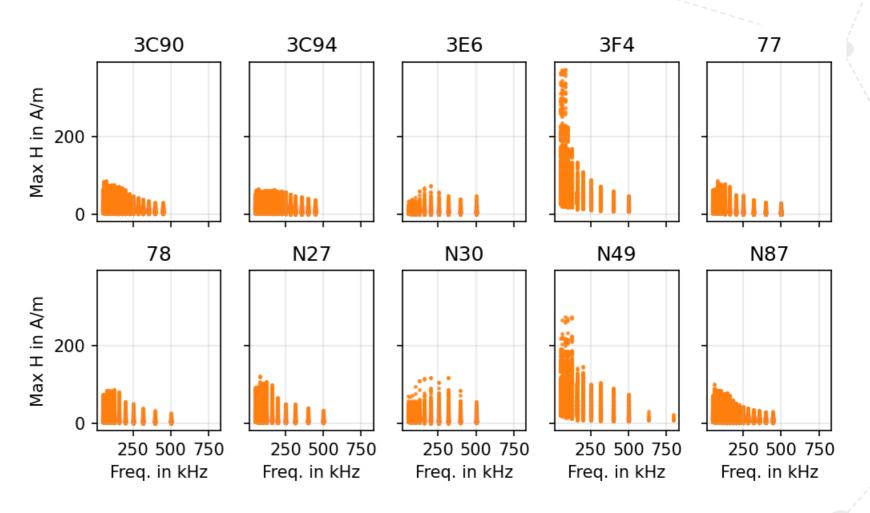


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# **Distributions (2)**

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 High max(H) only at low frequencies, for only two materials

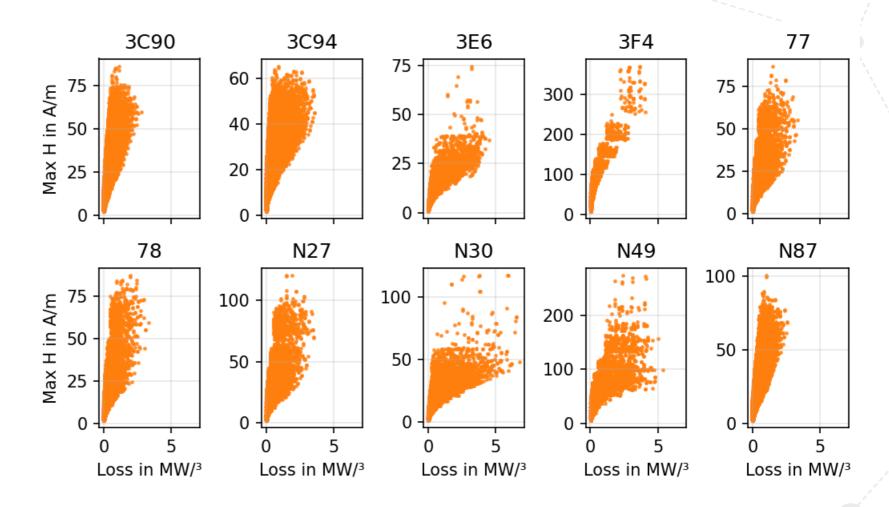


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# **Distributions (2)**

No perfect linear relationship

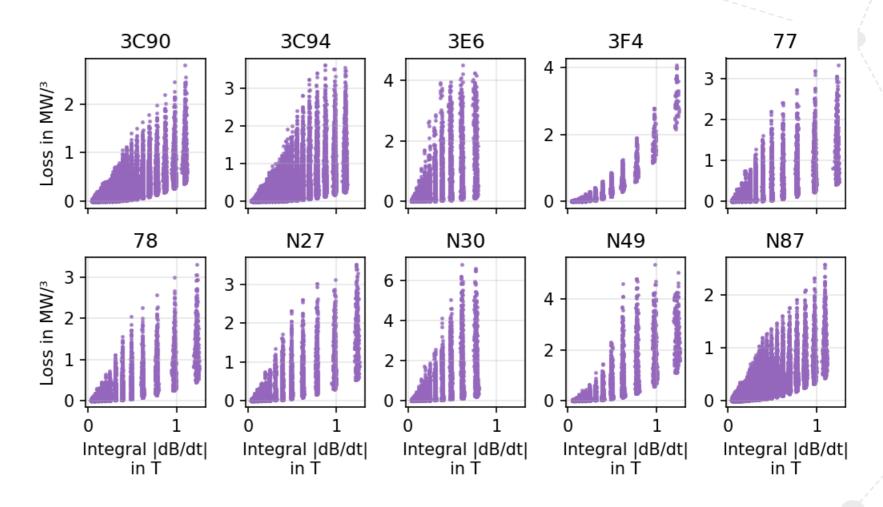


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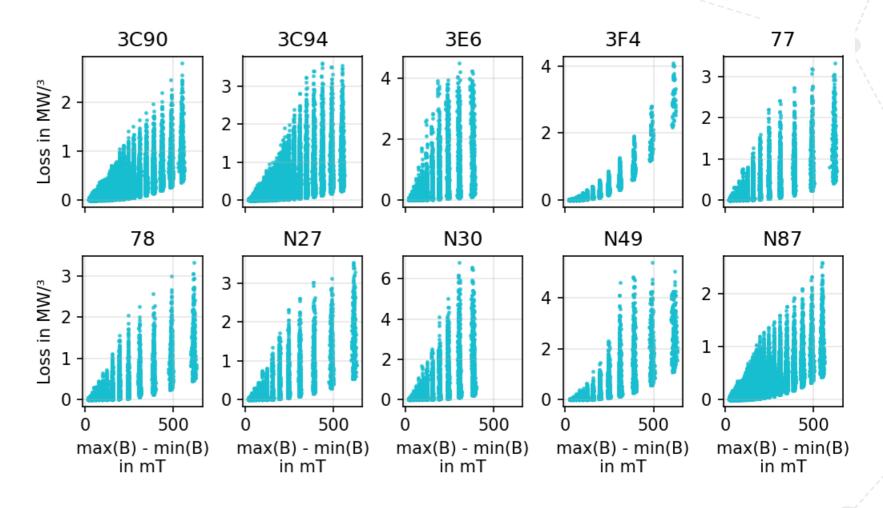
# **Distributions (2)**

- Some linear to exponential relationship
- High scatter



# **Distributions (2)**

• Similar to dB/dt

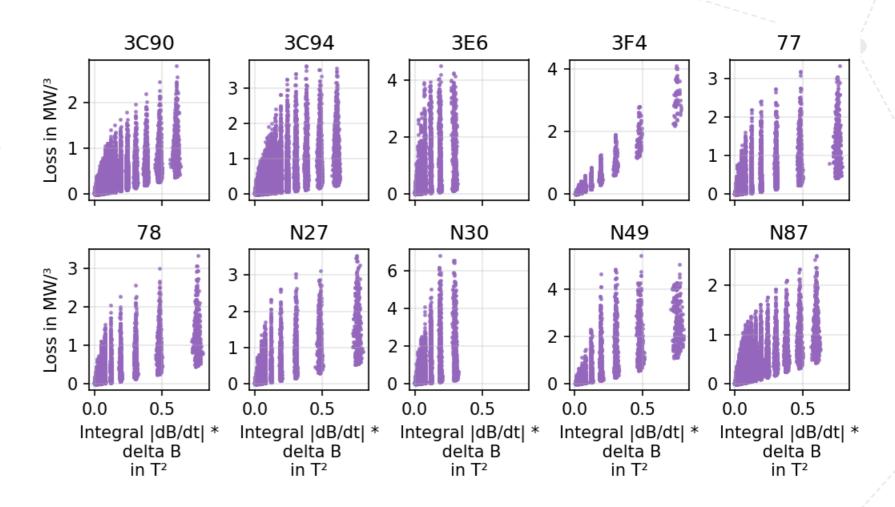


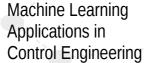


# **Distributions (2)**

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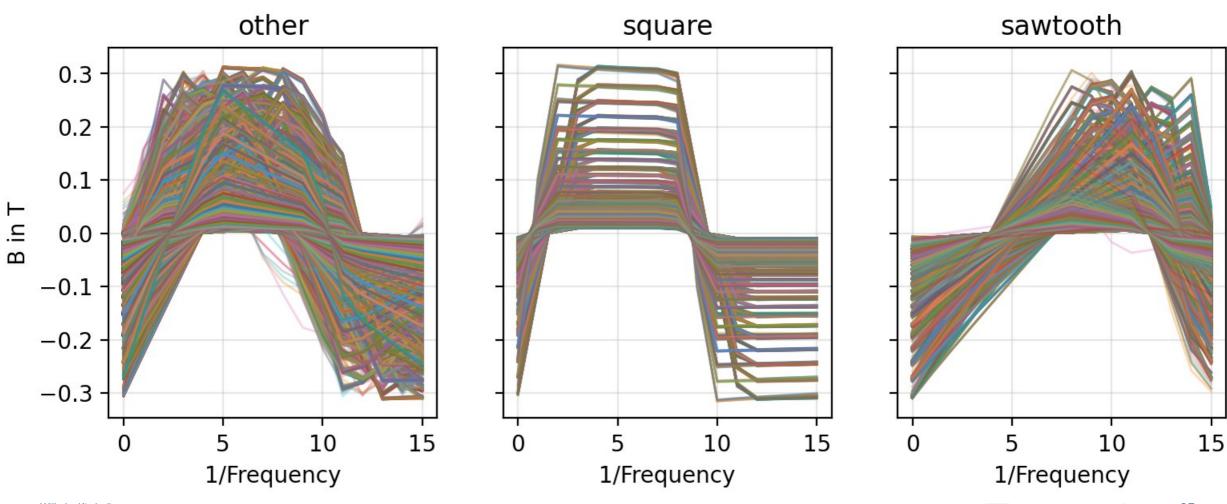
- Exponential relationship becomes more linear
- Scatter does not reduce







# Waveforms need to be estimated (here, template matching/ threshold-based)

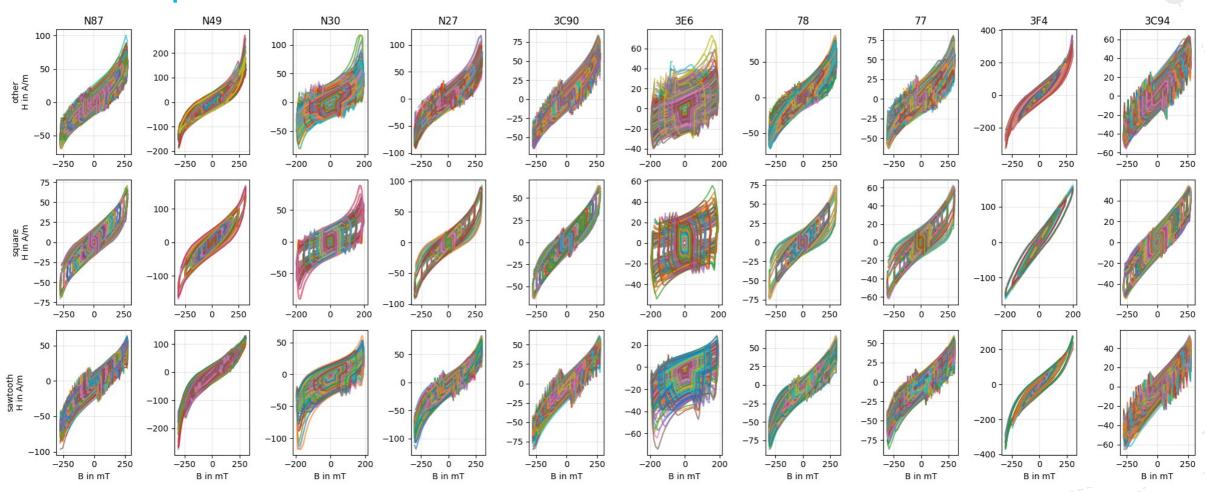


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# **B-H-Curves per waveform**

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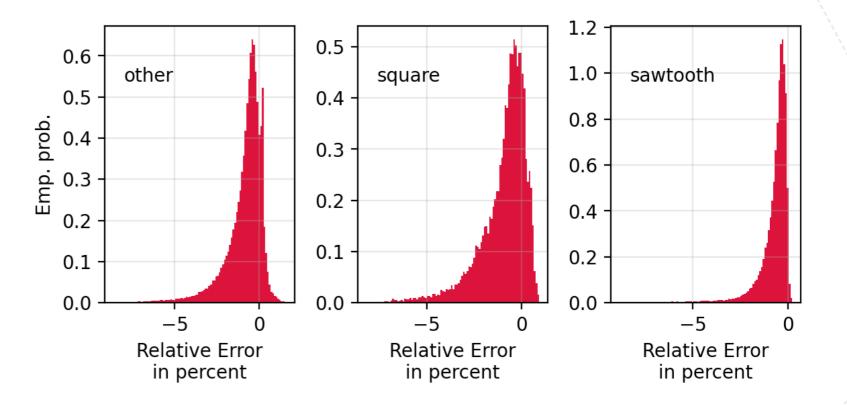
Square waveform has less distinct fins





#### **B-H-Area vs. Power loss**

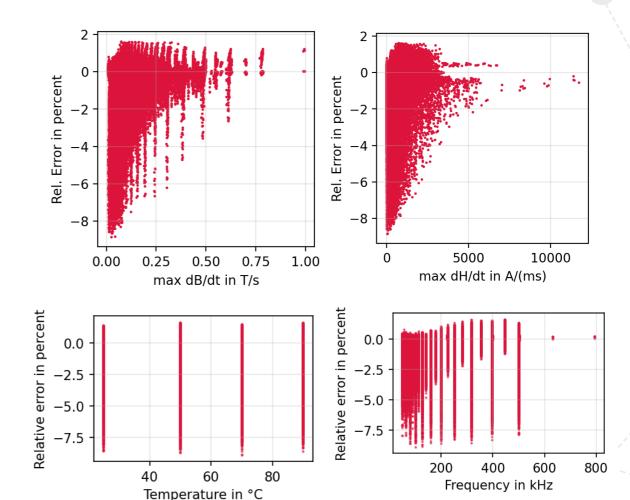
- Square waveform exhibits largest errors
- Sawtooth the smallest
- BH-Area tends to underestimate the loss
- Peaks are not at 0





#### **B-H-Area vs. Power loss**

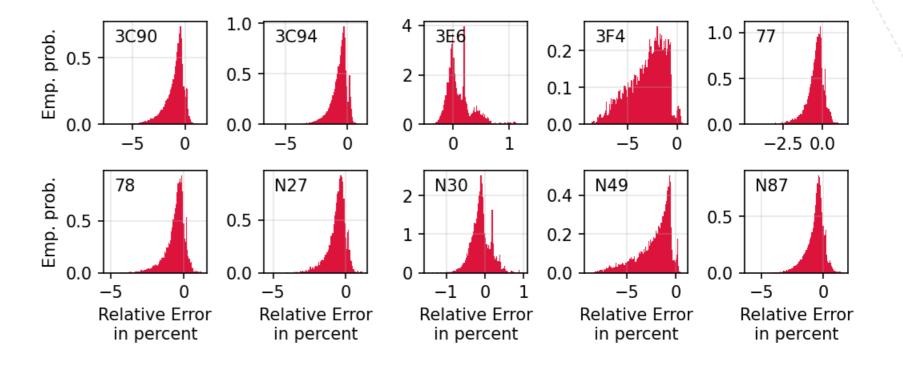
- Higher errors occur for small max(dB/dt) and max(dH/dt)
- Modes become visible
- Error has no correlation with temperature
- Error has two modes with frequency





# **B-H-Area vs. Power loss per material**

- 3E6, N30, and 77 show smallest errors
- 3F4 has the highest



https://git.uni-paderborn.de/lea-git/magnet-challenge-2023



#### Conclusion

- Estimating the H-curve is only a proxy of the loss at ~5% accuracy
- B-H curves look different for different materials (shape, max values, etc.)
- No strong linear or quadratic correlation evident yet (except max(B) → max(H))
- Modeling should consider
  - One model for all materials vs. one model per material
  - Approximating curves by subsampling (less data)
  - Merge Seq2Seq modeling B-curve → H-curve + correction with a regression model
  - Strong static modeling techniques such as Gradient Boosting Machines (XGBoost, LightGBM, CatBoost, etc.)
  - Initial value problem solvers forward/backward for estimation of H-curve

EDA

# Thank you



https://git.uni-paderborn.de/lea-git/magnet-challenge-2023

Wilhelm Kirchgässner