

# ML methods for superconducting magnets

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# Superconducting materials

- Super-conducting materials: materials that can achieve a zero-resistance state for particular values of temperature, magnetic field, and current density
- Quench: sudden and irreversible transition of a portion of conductor to resistive state in a superconducting magnet [SZ19]
- Due to large stored energy magnets can be damaged by quench (e.g. localized heating, high voltage).

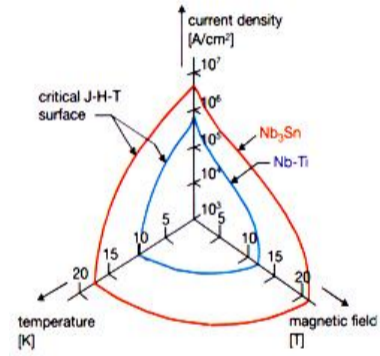


Figure 1: Superconducting region for different alloys magnets

# Magnet training

- Spontaneous quenches happen below the expected magnet performance
- Magnets training: it takes several quenches before reaching nominal performance
- Current is increased until quench happen (current ramp)
- More current ramps constitute a thermal cycle
- Magnets mentioned here are Canted-Cosine-Theta (CCT) subscale magnets
- Sub-scale magnets present lot of diagnostics, including quench antenna (QAs)



Figure 2: Example of QA on a subscale magnet

# Quench antenna

- Quench antennas provide information on events location and time
- QAs measure  $V = \sum_{i=1}^n \frac{d}{dt} (\int B dA_i)$
- A change in the voltage can occur for different reasons:
  - ▶ mechanical (epoxy cracks, friction, slipping)
  - ▶ electromagnetic (current redistribution, flux jumps)

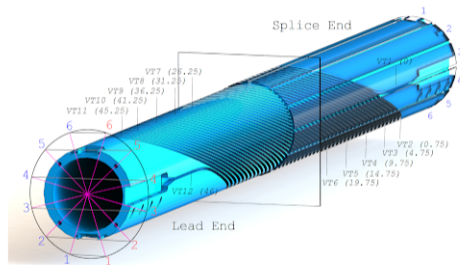


Figure 3: Model of QA on the CCT magnet at LBL

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# Global signal analysis

**Dataset** Data from two different CCT subscale magnets from LBNL:

- **sub2:** NHMFL mix-61 - baseline (standard) epoxy impregnated magnet
- **sub4:** CTD 701X - ultra tough (epoxy-non-epoxy) impregnated magnet

## Parameters tuning

- different threshold depending on the sampling frequency
- different filters for the different magnets

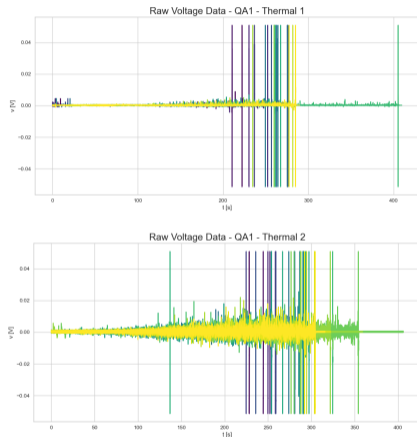


Figure 4: Raw data registered from QA1 for ramps in different thermal cycle

## Event detection - different antenna

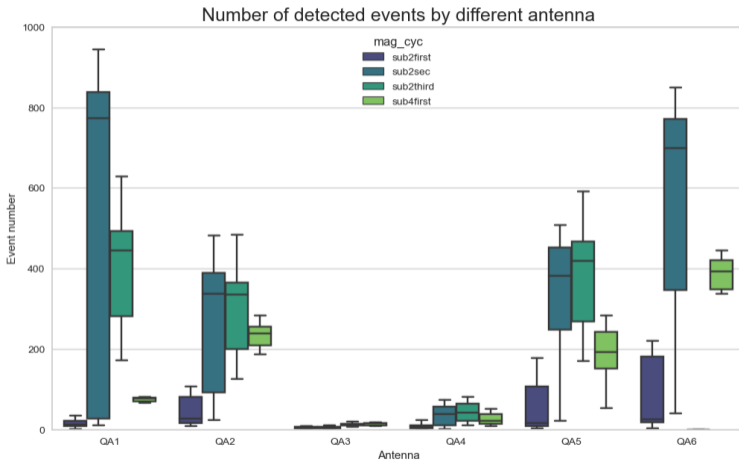


Figure 5: Number of events comprehensively detected by different antennas along the first, second and third thermal cycle for magnet `sub_2` and first thermal cycle for magnet `sub_4`



# Event Samples

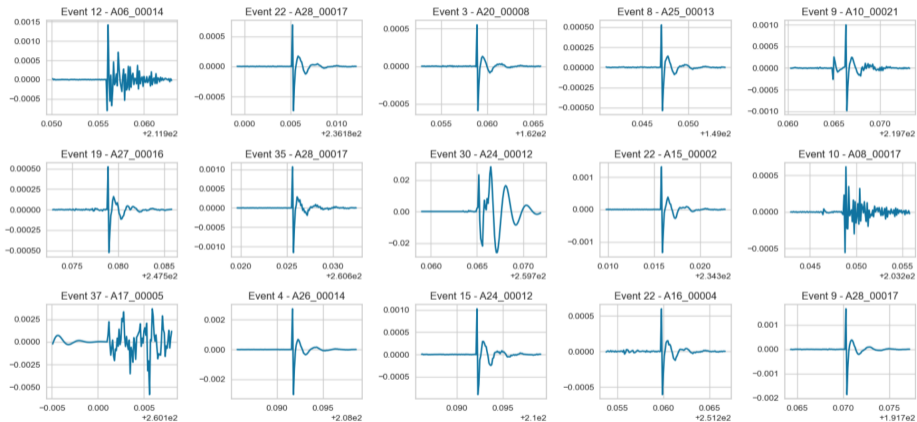


Figure 6: Samples of events from different ramp from the first thermal cycle - magnet sub\_2

# First Features Study

- The results presented here refers to data from QA1 antenna for the 23 current ramps of the 1st thermal cycle
- Considered three groups of features: **voltage** features, **signal** features and **frequency** features.
  - ▶ Voltage: max\_volt, min\_volt, norm, abs\_max, diff\_t\_Mm, full\_integral, arclen
  - ▶ Signal: time\_80, time\_50, time\_20
  - ▶ Frequency: occ\_1000, cwt\_lead, num\_peaks

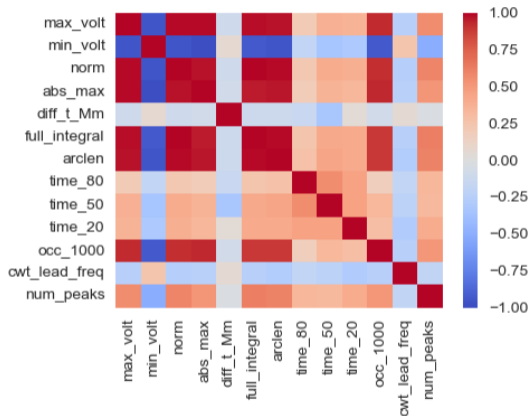


Figure 7: Heat-map of the correlation matrix for the mentioned features. We are looking for non-correlated features.

# Principal Component Analysis

- Principal Component Analysis (PCA) for dimensional reduction: all features are taken into consideration
- particularly important due to the data set size
- PCA retaining at least 90% of the information
- Explained Variance Ratio (EVR):
  - ▶ Comp\_1 : 0.650817
  - ▶ Comp\_2 : 0.140681
  - ▶ Comp\_3 : 0.094779
  - ▶ Comp\_4 : 0.046818

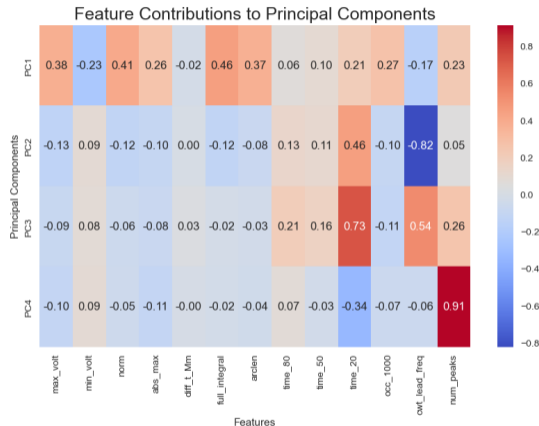


Figure 8: Heatmap representing features contributions to each PCA component

# K-means clustering

- Clustering performed over all events identified along 1 complete thermal cycle (sub2, thermal 1 - 23 ramps)
- Final choice for number of clusters determined using:
  - ▶ Silhouette Score: 4 clusters (0.748)
  - ▶ Number of negative silhouette score: 4 clusters (5 points)
  - ▶ K-Elbow: 4 clusters
  - ▶ WSS cross validation: 4 clusters (WSS score of 18.16)

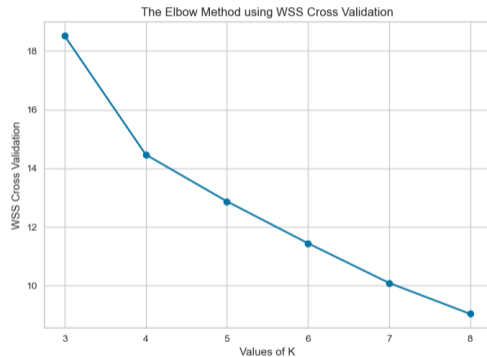


Figure 9: K-Elbow plot

# Clustering results

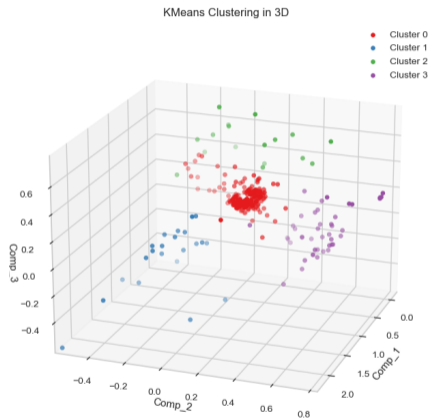
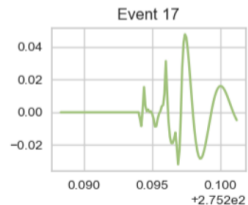
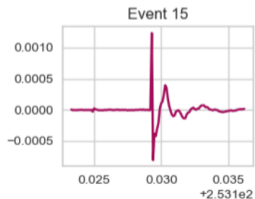
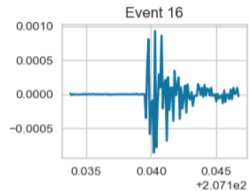
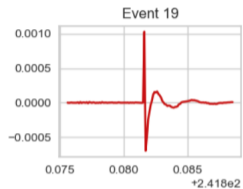


Figure 10: 3D cluster representation in PC space

# Event Samples from clustering

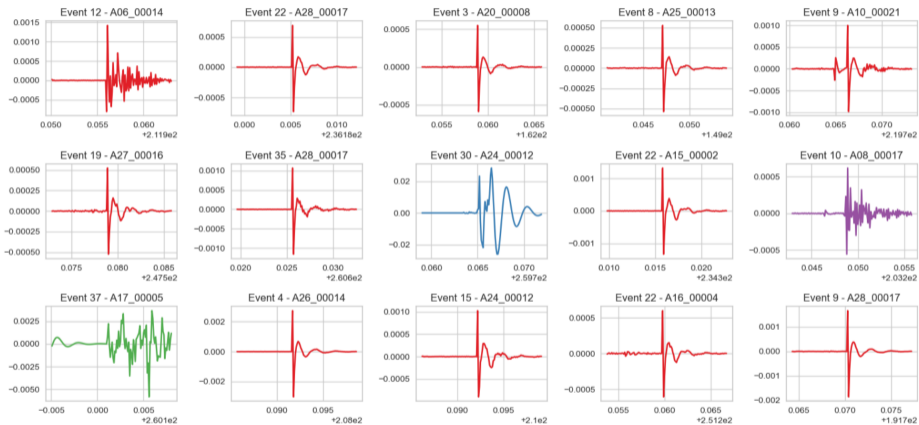


Figure 11: Samples of events from different ramp from the first thermal cycle - magnet sub 2 after clustering

# Feature boxplots

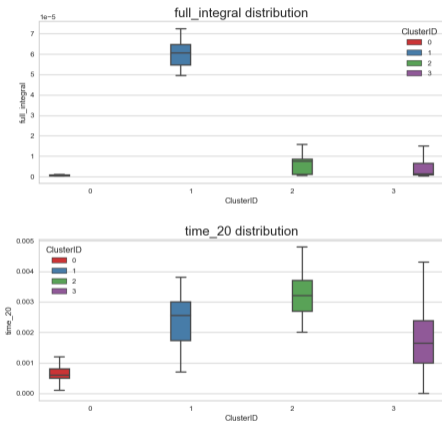


Figure 12: Boxplot of the integral value and of the time at which the 20% of the maximum voltage is reached in each cluster

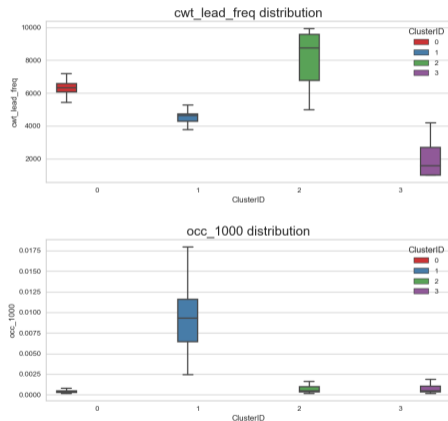


Figure 13: Boxplot of the lead frequency value and of the occurrence of 1kHz frequency in each cluster

## Feature distribution across clusters

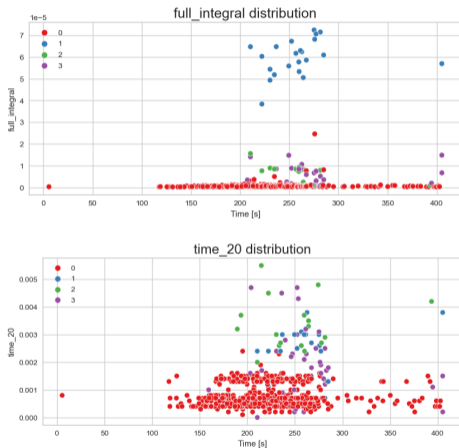


Figure 14: Integral value and reach of 20% of the maximum voltage distribution

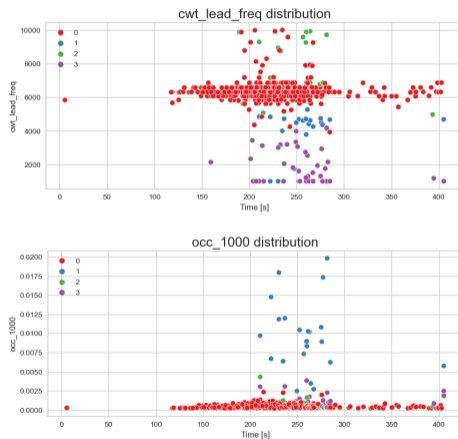


Figure 15: Lead frequency distribution in CWT analysis and occurrence of 1kHz frequency



# Cluster distribution in time

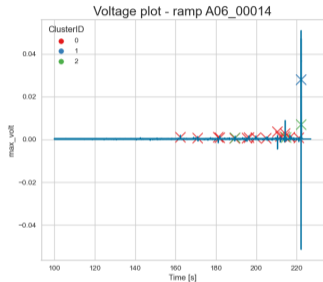


Figure 16: Example of events distribution for ramp A06\_00014 - thermal 1

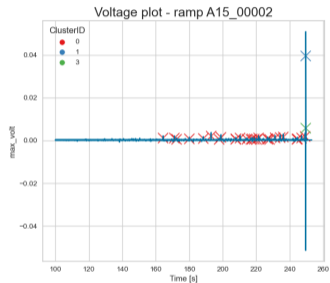


Figure 17: Example of events distribution for ramp A15\_00002 - thermal 1

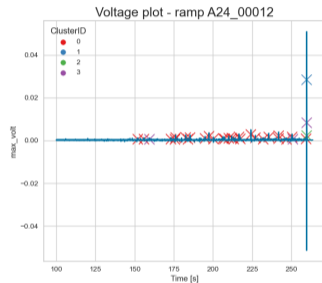


Figure 18: Example of events distribution for ramp A24\_00012 - thermal 1

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# Results

- Development of a tool that can consistently label events and detect similarity between events
- There is a very large increase in events in **sub2** after the thermal cycle, which needs to be better understood
- Remarkably similar events which has implications on the underlying disturbances better understanding is needed in the future
- The low number of events in QA3 and QA4 are likely due to the orientation of antennas with respect to the field direction
- **sub4** virgin ramp shown to have more events than the baseline magnet, as a result of the impregnation material mechanics

# What's next

- Physics:
  - ▶ More work is necessary to understand the variation of clusters across the magnet data
  - ▶ Try to locate the signal source
  - ▶ Compare properties and feature of the two magnets `sub2` and `sub4`
- Algorithm:
  - ▶ Using both acoustic and quench antenna data to improve results
  - ▶ Test different clustering techniques to improve efficiency and flexibility of the model
  - ▶ Improve hyper-parameter tuning depending on different data features
  - ▶ Try to implement an adaptive algorithm



Thank You!

# References I

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- [SZ19] Daniel Schoerling and Alexander V. Zlobin, eds. *Nb3Sn Accelerator Magnets*. Springer International Publishing, 2019. DOI: [10.1007/978-3-030-16118-7](https://doi.org/10.1007/978-3-030-16118-7). URL: <https://doi.org/10.1007/978-3-030-16118-7>.
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# Feature distribution along ramps

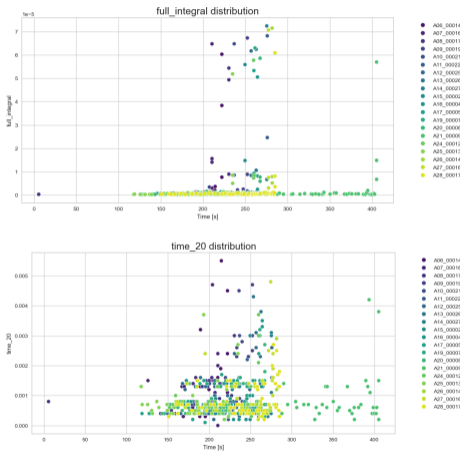


Figure 19: Example of feature distribution along different ramps

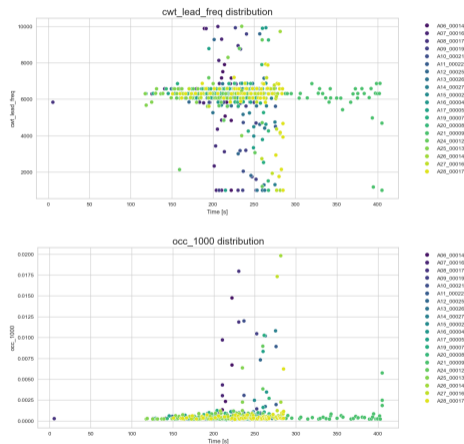


Figure 20: Example of feature distribution along different ramps

# Dataset details

Data from two different CCT subscale magnets from LBNL

**sub2:** standard epoxy impregnated magnet

- 3 different thermal cycles
- 3 different sampling frequency: 10 kHz, 20 kHz, 25 kHz respectively for the 1st, 2nd, 3rd thermal cycle
- data from all antennas (QA1 - QA6)

**sub4:** epoxy-non-epoxy impregnated magnet

- 1 thermal cycle
- sampling frequency: 25 kHz
- data coming from antenna QA1 - QA5



# Global signal analysis details

- different threshold depending on the sampling frequency
- different filters:
  - ▶ **sub2**: high pass filter with 4500 Hz cutoff frequency
  - ▶ **sub4**: band stop filter with 4000 - 5000 Hz cutoff frequency
- different thresholds:
  - ▶ 0.5 mV for 10 kHz and 20 kHz
  - ▶ 1 mV for 25 kHz sampling frequency

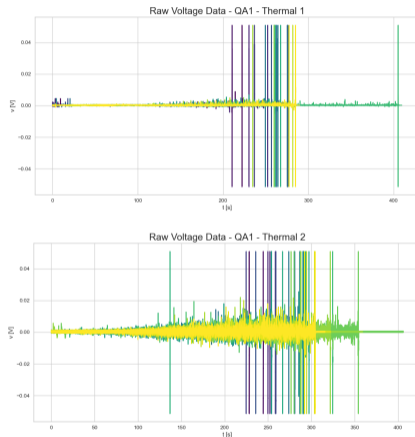


Figure 21: Raw data registered from QA1 for ramps in different thermal cycle