Analysis 00000

Features

Summary and next steps 00

Machine Learning methods for superconducting magnet applications

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Superconducting materials

Superconductor:

- Material which achieves superconductivity, a state of matter that has no electrical resistance
- Extremely important for accelerators, because they can generate strong magnetic fields which provide strong bending and focusing of the beam
- Requires very low temperatures: Fermilab best magnets need 1.9 K to provide 14.6 T





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Quench antennas

- Quench: sudden and irreversible transition of the superconductor into the normal-conducting state
 - After the quench, the energy stored in the magnet must be dissipated in order to protect the magnet
- Quench antennas: pick-up coil arrays sensitive to changes in the magnetic flux ⇒ provides quench identification and localization







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 $\underset{OO}{\text{Summary and next steps}}$

Voltage changes

- Quench antennas measure activity along the ramp, all the way to quench, at which point current is extracted from the magnet
 - Once the quench happens, a dump resistor is switched on to convert magnetic energy into thermal energy in the dump resistor



- Voltage spikes along the ramp caused by many possible reasons, some of which are:
 - Current redistribution within the cable
 - Frictional slipping of the cable
 - Vibration of the magnets
 - Epoxy cracks

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Aim of my work

- My job is to analyse the data provided by the quench antennas, finding the events prior to the quench, in order to find some specific features
- These features will then be fed to an unsupervised ML clustering algorithm in order to classify the events





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Filter			

 To eliminate the background noise (v = 4.54 kHz), a third-order Butterworth bandstop filter was implemented, allowing only frequencies lower than 4 kHz and higher than 5 kHz



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Integral

- To evaluate the magnetic flux received by the quench antennas, I evaluated the integral of the voltage
- The integral was performed using the trapezoid method on the voltage signal



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Event Selection

- Looking for events prior to the quench, $\nu_{samp} = 25 \text{ kHz}$
 - From a continuous waveform, trying to build windows around the selected events
 - Event selected when there are values above the positive threshold and below the negative threshold close to the positive peak



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Threshold analysis



Figure: Plot representing the events found using different thresholds, for both the raw signal and the filtered signal. The thresholds were selected between 1 mV and 4 mV



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Threshold analysis: number of events found

Threshold (mV)	Raw data	Bandstop Filter data	
1	147	95	
1.2	90	66	
1.4	67	51	
1.6	51	45	
1.8	36	35	
2	31	32	
2.2	27	25	
2.4	24	18	
2.6	22	17	
2.8	20	14	
3	17	14	
3.2	13	13	
3.4	13	11	
3.6	12	10	
3.8	11	7	
4	10	6	

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Features matrix

- Prior to performing ML techniques, I first need to identify some features which will be used for the clustering
- Features extracted from each event, both for raw and filtered signal
- The features will then be added to a matrix, which will then be fed to an unsupervised clustering algorithm to identify the representative events



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Description of features

- I divided the features in four main categories:
 - 1 Voltage signal features
 - Maximum and minimum value of the voltage
 - Norm of the voltage array
 - 2 Integrated signal (magnetic flux) features :
 - Maximum and minimum value of the integrated voltage
 - Norm of the integrated voltage array
 - Definite integral of the voltage
 - **3** Signal shape:
 - Width and tail of the signal
 - Signal length
 - 4 Fast Fourier Transform (FFT) features:
 - Peak frequency
 - Distribution of frequencies obtained with FFT analysis



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Plots displaying some features, both raw and filtered



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ML for Superconducting magnet applications

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Summary and next steps

- Goal: use Machine Learning to identify and learn about the disturbances in high field superconducting accelerator magnets
- Achieved so far:
 - **1** Build a routine to automatically extract windowed events from continuous data
 - 2 Analysis of windowed events for signal characteristics, signal shape and frequency distribution
- Next steps:
 - **1** Finalize FFT analysis on the signal
 - **2** Perform unsupervised *k*-means clustering on feature matrix
 - 3 Implement a Neural Network to identify the events and check prediction power on more data



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Thank you for the attention



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