



Intelliquench – Real Time detection of magnet quenches in superconducting radio frequency magnets.

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SIST Project Introduction

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**U.S. MAGNET
DEVELOPMENT
PROGRAM**

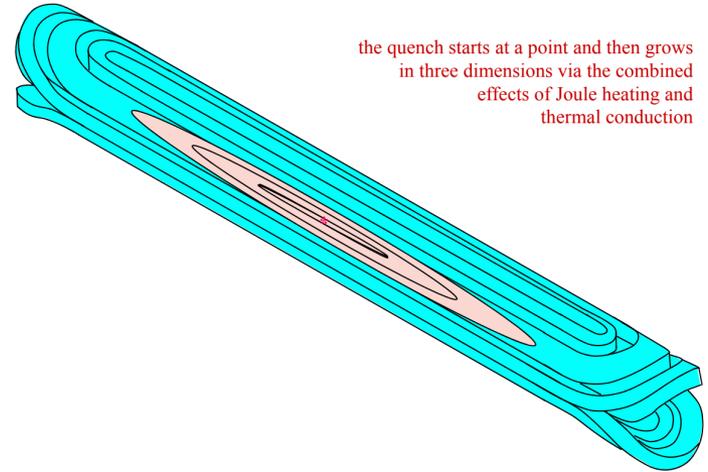
Outline

1. What are magnet quenches?
2. Consequences of magnet quenches.
3. Acoustic sensors (small microphones).
4. Deep Neural Network for monitoring acoustic signals to detect quenches.
5. Summary & Outlook

Magnet quenches

- Superconducting accelerator magnets must operate at **very low temperatures** to maintain superconductivity (**no resistance**).
- Due to several reasons (mechanical errors, conductor motion, ...), a **specific spot** in the magnet **heats up**.
- This causes the magnet to become **resistive**, and with **huge amount of current** pumping through, it can be catastrophic.

Growth of the resistive zone



Wilson et al. Superconducting magnets for accelerators.

r/CatastrophicFailure



In **2008**, magnet quench occurred in **100 magnets at the LHC at CERN**, leading to a loss of approximately **six tonnes** of liquid helium.

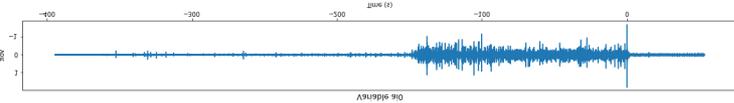


The escaping vapour expanded with **explosive force**, damaging a total of 53 superconducting magnets (each costs **several millions dollar.**)

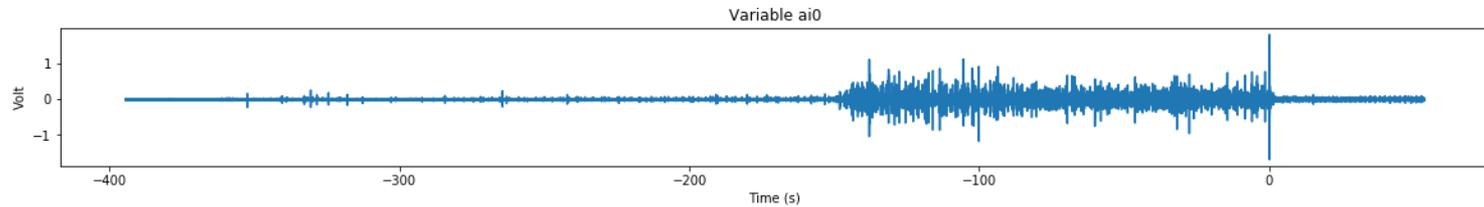


Acoustic sensors

- We placed **5 tiny microphones** around the magnet to detect **abnormal sound signatures**.



Deep Neural Network to detect anomaly in the signal.



Deep Neural Network



Abnormal sound signals?

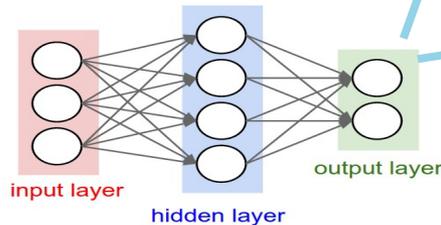
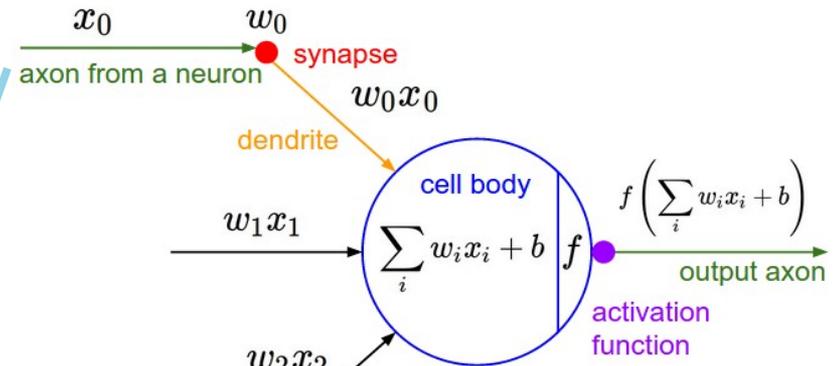
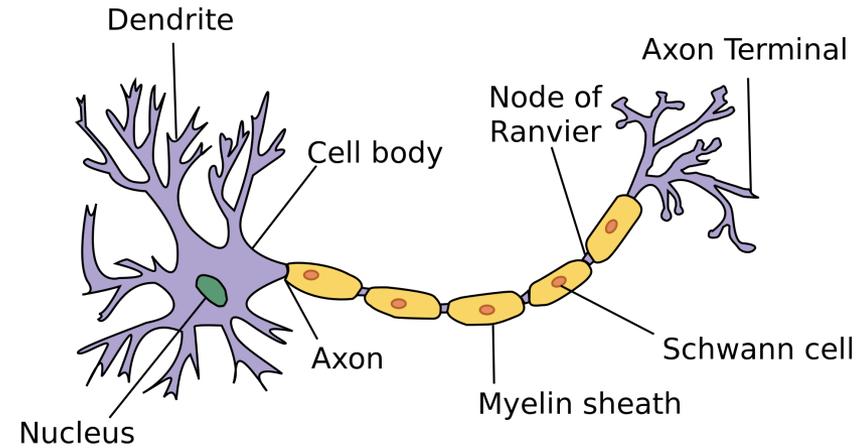
Summary & Outlook

1. Magnet quenches **are expensive**.
2. We are using Deep Neural Network to **detect anomaly sound signals**, which hopefully enable us to **predict the quench**.

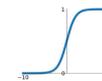
Back-ups

Deep Neural Networks

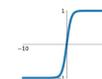
- Fully connected architecture
- Each **input** multiplied by a **weight**.
- **Weighted values** are summed, **Bias** is added.
- Non-linear **activation function** is applied
- Trained by varying the **parameters** to minimize a loss function (quantifies how many mistakes the network makes)



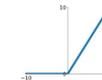
Sigmoid
 $\sigma(x) = \frac{1}{1+e^{-x}}$



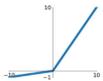
tanh
 $\tanh(x)$



ReLU
 $\max(0, x)$



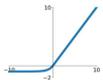
Leaky ReLU
 $\max(0.1x, x)$



Maxout
 $\max(w_1^T x + b_1, w_2^T x + b_2)$

ELU

$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$



DNN Auto-encoder

Deep neural network as auto-encoder

