

# My Ph.D.: Interpretable Fault Prediction

#### Ph.D. finish line: 19th of January

Main contribution:

- Breakdown prediction in CLIC RF cavities
	- Field emitted current following an initial breakdown is related to the probability of another breakdown occurring
	- C. Obermair et al., "Explainable Machine Learning for Breakdown Prediction in High Gradient RF Cavities", PRAB, 2022
- Interpreting ML models for fault prediction
	- Novel method for explaining fault predictions with ML, evaluation with 75 people from CERN and TU Graz
	- C. Obermair et al., "Example or Prototype? Learning Concept-Based Explanations in Time-Series", PMLR, 2022
- Interpretable Anomaly Detection in the LHC Main Dipole Circuit
	- Understand normal behavior and detect non-normal behavior in voltage measured at the diode after a FPA
	- C. Obermair et al., "Interpretable Anomaly Detection in the LHC Main Dipole Circuits with Non-negative Matrix Factorization", to be submitted to IEEE, 2024







### Interpretable Anomaly Detection in the LHC Main Dipole Circuit with Non-Negative Matrix Factorization

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Goal: Define and understand normal behavior, detect abnormal behavior of the main dipole (RB) circuit

#### Approach:

- 1. Extract frequencies in data  $\rightarrow$  Fast Fourier transform (FFT)
- 2. Group expected frequencies that occur together into components → Non-Negative Matrix Factorization (NMF)
	- a) Components help to understand normal behavior  $\rightarrow$  Causal Discovery
	- b) Deviations help to detect **abnormal** behavior  $\rightarrow$  Outlier detection







#### Signal:

• U\_diode from nQPS in PM

#### Region:

• Plateaus after energy extraction  $\rightarrow$  Similar to transient measurement

#### Period:

• 2018, Quench + Snapshot data

#### Data size:

• 731 events x 154 magnets x 400 samples (0.375s)





# Fast Fourier transform

Extract frequencies in data





### Fast Fourier transform

Example signal: B21R3 on 2021-04-18 08:44:17

Preprocessing necessary to minimize spectral leakage:

- Subtract offset
- Multiplication with window







![](_page_7_Figure_1.jpeg)

![](_page_7_Figure_2.jpeg)

![](_page_7_Picture_3.jpeg)

# Non-Negative Matrix Factorization (NMF)

Group expected frequencies that occur together into components

- a) Components help to understand normal behavior  $\rightarrow$  Causal Discovery
- b) Deviations help to detect **abnormal** behavior  $\rightarrow$  Outlier detection

![](_page_8_Figure_4.jpeg)

![](_page_8_Picture_5.jpeg)

<https://www.nature.com/articles/44565>

<https://proceedings.neurips.cc/paper/2000/file/f9d1152547c0bde01830b7e8bd60024c-Paper.pdf>

![](_page_9_Figure_0.jpeg)

Loss:  $\sum_{k} (|X[k]| - \big| \widehat{X}[k] \big| \big)^2$ 

How to define  $W_k$  &  $h_k$ ?

Grāz

- 1. Manually define  $r$
- 2. Initialize  $W_k$  &  $h_k$  randomly

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3. Adjust  $W_k$  &  $h_k$  iterativly until loss over all signals (763 \* 154 \* 2= 235 004) is minimal

![](_page_9_Picture_6.jpeg)

Example signal: B21R3 on 2021-04-18 08:44:17

### NMF Components

![](_page_10_Figure_1.jpeg)

![](_page_11_Figure_0.jpeg)

# Causal Discovery

Understand normal behavior

![](_page_12_Figure_2.jpeg)

![](_page_12_Picture_3.jpeg)

### Causal Discovery

Frequency components along all magnets in event

![](_page_13_Figure_3.jpeg)

![](_page_13_Figure_4.jpeg)

# Outlier Detection

Detect abnormal behavior

![](_page_14_Figure_2.jpeg)

![](_page_14_Picture_3.jpeg)

## Outlier Detection

Components state frequencies, expected to occur Normal event: Reconstruction with components possible (low loss) Outlier event: Unexpected frequencies occur (high loss)

#### How to find an outlier:

- 1. Calculate NMF loss for each event (763)
- 2. Fit gamma distribution to loss
- 3. Calculate p value for each event (763)
	- $\rightarrow$  probability of obtaining results at least as extreme as the observed

![](_page_15_Figure_7.jpeg)

![](_page_15_Picture_8.jpeg)

## **Outliers**

Goal: Find outliers robust to assumptions

- Result shows boxplot of 280 different combinations of assumptions
- All outliers occur during  $1^{st}$  EE Plateau  $\rightarrow$  closer in time to quench

![](_page_16_Figure_4.jpeg)

![](_page_16_Picture_5.jpeg)

![](_page_16_Picture_6.jpeg)

## Conclusion

#### Causal Discovery:

- Detection of "normal" frequency components with Non-Negative Matrix Factorization
	- $\rightarrow$  Depending on the quench position, a typical FPM would look like this:

#### Outlier Detection:

- Outliers are events which cannot be composed out of "normal" frequency components
- Ongoing additional measurements, additional safety measures, possible replacement

![](_page_17_Figure_7.jpeg)

![](_page_17_Picture_8.jpeg)

![](_page_18_Picture_0.jpeg)

Understand where **experts** can profit from data analysis!

#### Ongoing ML projects:

- U diode NXCALS data
	- Investigation of secondary quenches
- [mb-feature-classification](https://gitlab.cern.ch/machine-protection/mb-feature-classification)
	- Make decisions: Gather RB machine parameters and analysis results
	- Find correlations with ML
- UQS0 signal classification
	- Classify ~35000 UQS0 signals from snapshot FPAs similar as experts
	- 80% of signals are classified similarly
	- In the remaining 20%, ML was right in 90%
- SOH prediction in capacitor banks
	- Assisting Timm Baumann SY/EPC

![](_page_18_Picture_14.jpeg)

![](_page_18_Figure_16.jpeg)

![](_page_19_Picture_0.jpeg)

# Backup Slides

![](_page_20_Picture_1.jpeg)

## Aliasing

High frequency components could potentially cause aliasing in results. Anti-aliasing filters in the nQPS crates:

- Two 1<sup>st</sup> order lowpass filters with 1.5 kHz and 1 kHz cutoff frequency\*
- Sampling frequency of nQPS crates: 1068 Hz

![](_page_21_Figure_4.jpeg)

![](_page_21_Figure_5.jpeg)

![](_page_21_Picture_6.jpeg)

Aliasing Examples

# Fast Fourier transform

![](_page_22_Picture_1.jpeg)

#### **FFT** Fast Fourier transform

 $\rightarrow$  The FFT is an algorithm to calculate the discrete Fourier transform (DFT). The DFT is defined as:

![](_page_23_Figure_2.jpeg)

![](_page_23_Figure_3.jpeg)

![](_page_23_Picture_4.jpeg)

![](_page_24_Picture_0.jpeg)

 $\rightarrow$  Avoid smearing

![](_page_24_Figure_2.jpeg)

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 $1.0$ 

 $0.8$ 

 $0.6$  $0.4 0.2 0.0$ 

 $1.0$ 

 $0.8$  $0.6 -$  - hanning

- bartlett

 $0.25$   $0.30$   $0.35$   $0.40$   $0.45$   $0.50$   $0.55$ <br>Time / s

![](_page_25_Picture_0.jpeg)

→ Smearing of DC component interferes with low frequency component

#### $x[n] = 2V + 2V * sin(2\pi 3Hz * n - 90^\circ) + sin(2\pi 20Hz * n)$

![](_page_25_Figure_3.jpeg)

![](_page_25_Picture_4.jpeg)

# Backwards Path

#### Signal: B21R3 on 2021-04-18 08:44:17

![](_page_26_Figure_2.jpeg)

 $\overline{5}$  -0.001  $-0.002$  $-0.003$  $-0.004$  $0.25$ 

 $0.30$ 

 $0.35$ 

 $0.40$ 

Time / s

 $0.45$ 

 $0.50$  $0.55$ 

 $\longrightarrow x^*[n]$ 

 $0.50 0.55$ 

 $\rightarrow x^*[n]$ 

 $\hat{x}_{FFT}^*[n]$ 

**FFT** 

 $X[k]=\sum_{n=0}^{N-1}x^*[n]e^{-i2\pi nk/N}$ 

 $|X[k]|, \varphi$ 

**IFFT** 

![](_page_26_Figure_3.jpeg)

![](_page_26_Picture_4.jpeg)

# Non-Negative Matrix Factorization (NMF)

![](_page_27_Picture_1.jpeg)

<https://www.nature.com/articles/44565> <https://proceedings.neurips.cc/paper/2000/file/f9d1152547c0bde01830b7e8bd60024c-Paper.pdf>

## Objective Function

#### $V \approx WH$

 $m$  ... number of  $i$  events  $*$  positions (560 x 154)  $n$  ... number of frequencies (0-360Hz)  $r$  ... number of components (1-5)  $V \in \mathbb{R}^{n \times m}_+$  ... reconstructed event at position  $W \in \mathbb{R}_+^{n \times r}$  ... components  $H \in \mathbb{R}_+^{r \times m}$ ... presence of components

NMF Objective Function:  $\min_{W,H} f(W,H) \equiv \frac{1}{2}$  $\frac{1}{2}||V-WH||_F^2$ , s.t.  $W, H \ge 0$ 

![](_page_28_Figure_4.jpeg)

![](_page_28_Picture_5.jpeg)

## Backwards Path

Signal: B21R3 on 2021-04-18 08:44:17

![](_page_29_Figure_2.jpeg)

# **Causal Discovery**

![](_page_30_Picture_1.jpeg)

## El. Vs Phys. Position

![](_page_31_Figure_1.jpeg)

![](_page_31_Figure_2.jpeg)

![](_page_31_Picture_3.jpeg)

### U-diode frequencies

![](_page_32_Figure_1.jpeg)

![](_page_32_Picture_2.jpeg)

![](_page_33_Picture_0.jpeg)

![](_page_33_Picture_17.jpeg)

![](_page_33_Picture_3.jpeg)

![](_page_34_Figure_1.jpeg)

## 3Hz Examples

![](_page_35_Figure_1.jpeg)

Quenched Magnet: C32L2 (Manufacturer 1) RB\_RB.A12\_1619462088820000000\_2021-04-26

U\_Diode @ 1st EE plateau

![](_page_35_Figure_3.jpeg)

![](_page_35_Figure_4.jpeg)

Quenched Magnet: C14L2 (Manufacturer 1) RB\_RB.A12\_1619935955860000000\_2021-05-02

![](_page_35_Figure_6.jpeg)

![](_page_35_Figure_8.jpeg)

Quenched Magnet: B33R4 (Manufacturer 1) RB\_RB.A45\_1620232873800000000\_2021-05-05

![](_page_35_Figure_10.jpeg)

![](_page_35_Picture_11.jpeg)

 $0.35$ 

 $0.40$ 

 $0.45$ 

 $0.50$ 

Time / s

 $0.55$ 

 $0.60$ 

0.65

 $-4.6$ 

 $-4.8$ 

 $-5.0$ 

 $-5.2$ 

 $-5.4$ 

age / V

![](_page_36_Figure_1.jpeg)

![](_page_37_Figure_1.jpeg)

![](_page_38_Figure_1.jpeg)

![](_page_38_Picture_2.jpeg)

![](_page_39_Figure_1.jpeg)

![](_page_39_Picture_2.jpeg)

![](_page_40_Figure_1.jpeg)

It actually starts 1 QPS crate after the EE switches

![](_page_40_Picture_3.jpeg)

![](_page_41_Figure_1.jpeg)

![](_page_42_Figure_1.jpeg)

![](_page_42_Picture_2.jpeg)

![](_page_43_Figure_1.jpeg)

![](_page_43_Picture_2.jpeg)

# Outlier Detection

![](_page_44_Picture_1.jpeg)

### Assumptions

#### Assumptions for this plot:

- Linear detrend
- Hamming window
- 4 components
- Frobenius distance

![](_page_45_Figure_6.jpeg)

#### 1. Preprocessing:

- 1. Degree of detrend:
	- 1. 0 Offset
	- 2. 1 Linear Trend
- 2. Window multiplication:
	- 1. none (best reconstruction, high smearing)
	- 2. hanning (lowest smearing, no reconstruction)
	- 3. hamming (low smearing , good reconstruction)
	- 4. barlett
	- 5. blackman
	- 6. flattop (high smearing, accurate amplitude)
	- 7. tukey

#### 2. NMF:

- 1. n components (2-12)
- 2. Distance measure\*:
	- 1. Frobenius (Eu)

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2. Kullback-Leibler (KL)

\* https://ieeexplore.ieee.org/stamp/stamp.jsp?arnumber=6410389

### 2038 - RB.A78 - B28L8 - Intercoil short

Event with B28L8 quench before:

2021-03-28: "normal" at EE plateaus  $\bullet$ 

FPA identifier: RB RB.A78 1619330143440000000 Date: 2021-04-25 07:55:43.418000 Max. Current: 11588.0 A

El. Position Primary Primary quench position: 126 Fast secondary quench: []

![](_page_46_Figure_5.jpeg)

El. Position

![](_page_46_Figure_6.jpeg)

![](_page_46_Figure_7.jpeg)

![](_page_46_Picture_8.jpeg)

### 1225 - RB.A45 - C17L5

Events with C17L5 quench before:

- 2021-05-07: "normal" at EE plateaus
- 2021-05-07: "normal" at EE plateaus

FPA identifier: RB RB.A45 1620797547820000000 Date: 2021-05-12 07:32:27.799000 Max. Current: 11701.0 A

 $40$ 

60

 $20$ 

El. Position Primary Primary quench position: 90 Fast secondary quench: []

![](_page_47_Figure_6.jpeg)

![](_page_47_Figure_7.jpeg)

U\_Diode Signals

80

El. Position

100

 $120$ 

140

![](_page_47_Picture_8.jpeg)

### 1146 - RB.A34 - A32L4

Events with A32L4 quench before:

- 2021-04-04: "normal" at EE plateaus
- 2021-04-14: 3 fast sec. quenches

FPA identifier: RB RB.A34 1620323722320000000 Date: 2021-05-06 19:55:22.295000 Max. Current: 11950.0 A

40

 $20$ 

El. Position Primary Primary quench position: 120 Fast secondary quench: []

![](_page_48_Figure_6.jpeg)

![](_page_48_Figure_7.jpeg)

![](_page_48_Figure_8.jpeg)

**U\_Diode Signals** 

80

El. Position

100

120

60

 $140$ 

![](_page_48_Figure_9.jpeg)

![](_page_48_Picture_10.jpeg)

### 1291 - RB.A12 - B11L2

#### No B11L2 quench before

FPA identifier: RB\_RB.A12\_1621014819920000000 Date: 2021-05-14 19:53:39.901000 Max. Current: 11751.0 A

El. Position Primary Primary quench position: 151 Fast secondary quench: []

![](_page_49_Figure_4.jpeg)

![](_page_49_Figure_5.jpeg)

![](_page_49_Figure_6.jpeg)

![](_page_49_Figure_7.jpeg)

![](_page_49_Picture_8.jpeg)

### 2421 - RB.A34 – B28R3

No B28R3 quench before

Most likely scenario of noise from simulations:

- Partially emerging resistor, in parallel to diode
- Degraded diode contact?

Date: 2021-04-20 07:28:30.924000 Max. Current: 11786.3 A U Diode Signals

FPA identifier: RB RB.A34 1618896510960000000

![](_page_50_Figure_6.jpeg)

El. Position Primary Primary quench position: 106 Fast secondary quench: ['49@198ms']

![](_page_50_Figure_8.jpeg)

![](_page_50_Figure_9.jpeg)

![](_page_50_Figure_10.jpeg)

![](_page_50_Figure_11.jpeg)

![](_page_50_Picture_12.jpeg)