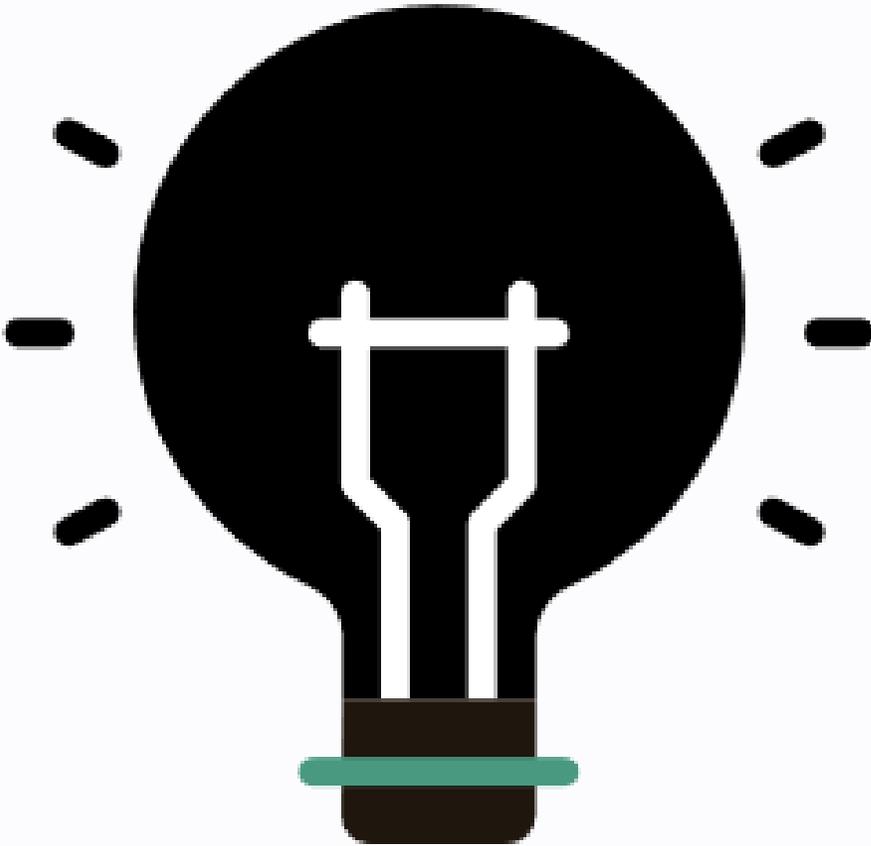


# Kourtogram

Mostafavi, Atabak  
Friedmann, Andreas

# ● Introduction



**Kourtoqram**

**Limitation**

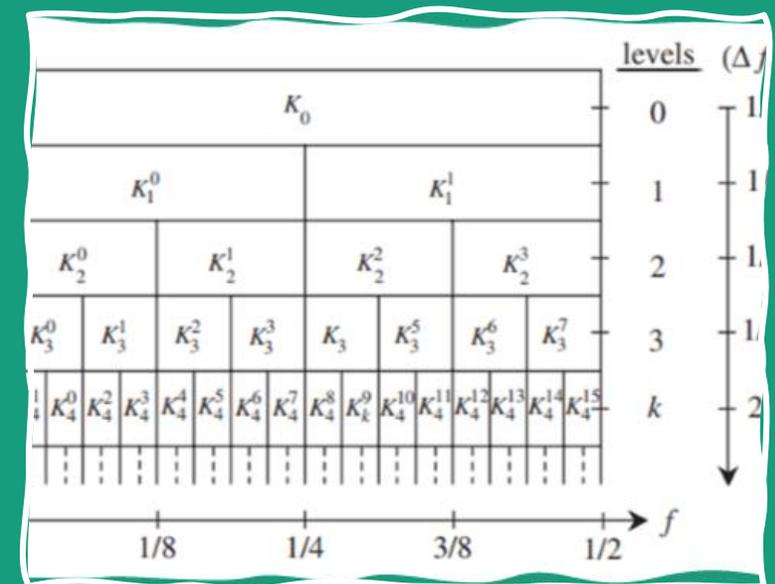
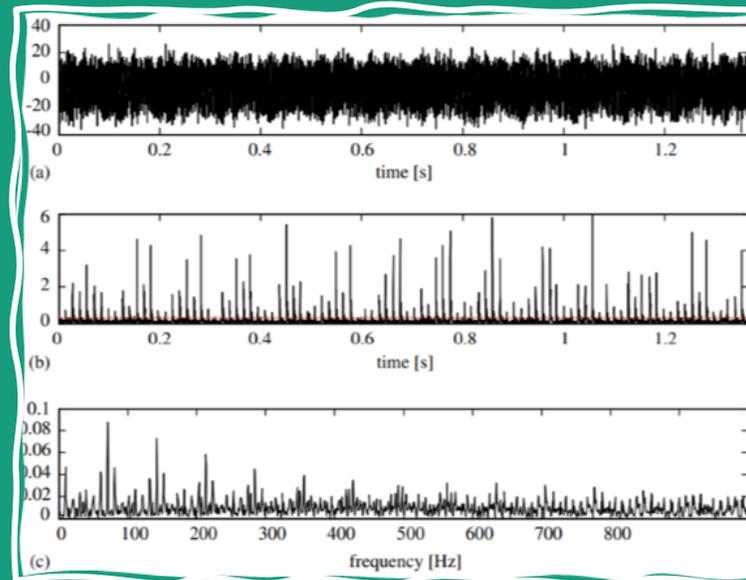
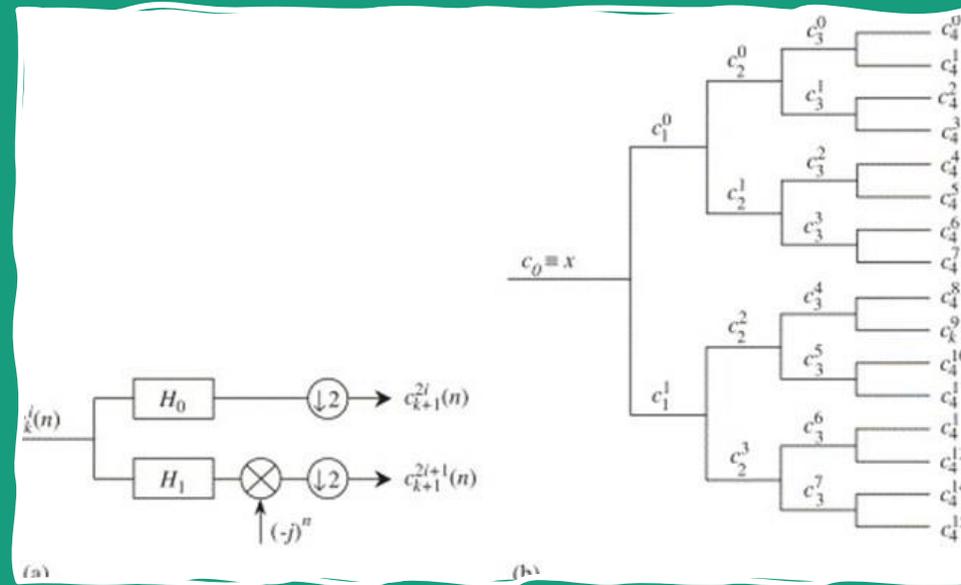
**Solutions**

**Case study**

# Kourtogram

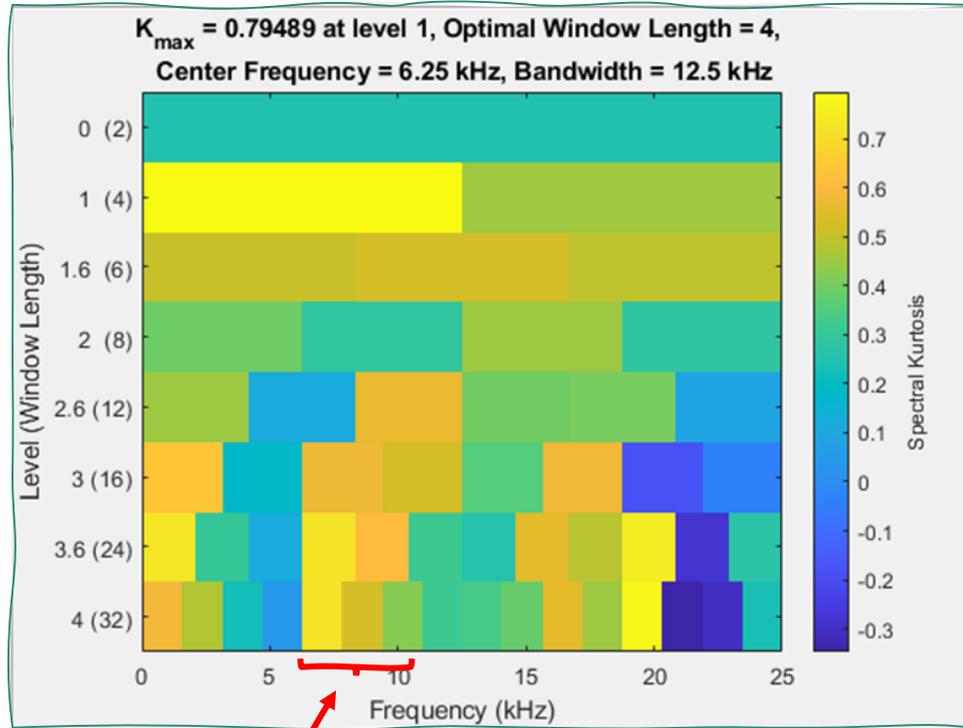
- Purpose
  - Transient detection
  - Non stationarity
- Why not Global Kurtosis
- Algorithm
- Filter Window detection
- Application on bearing and gearbox

**Source:** Fast computation of the kurtogram for the detection of transient faults , Antoni(2006)



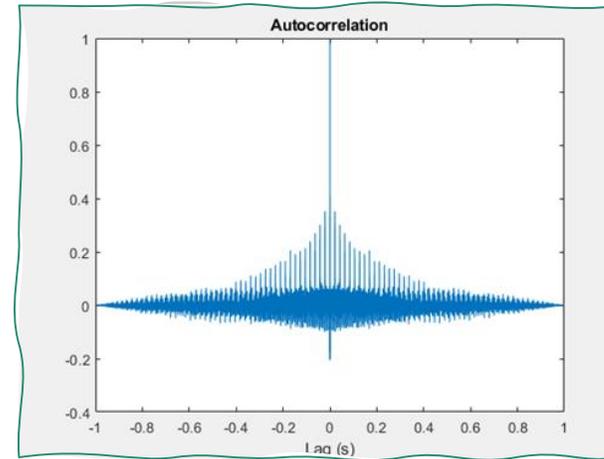
# Ambiguous Kourtogram (first case)

Excessive filter bandwidth (Accelerometer Data from bearing with outer race fault (MAFAULDA). RPM= 48 rev/s)

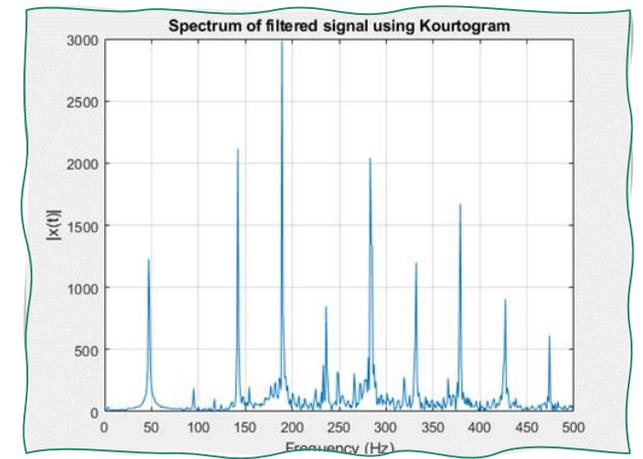


Damage Bandwidth

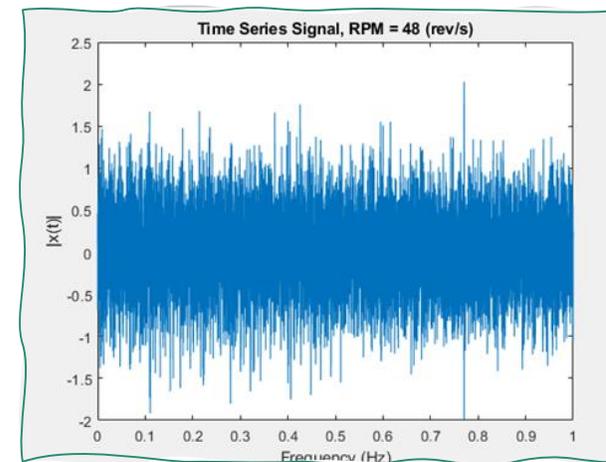
Autocorrelation



Filtered Spectra using Kourtogram

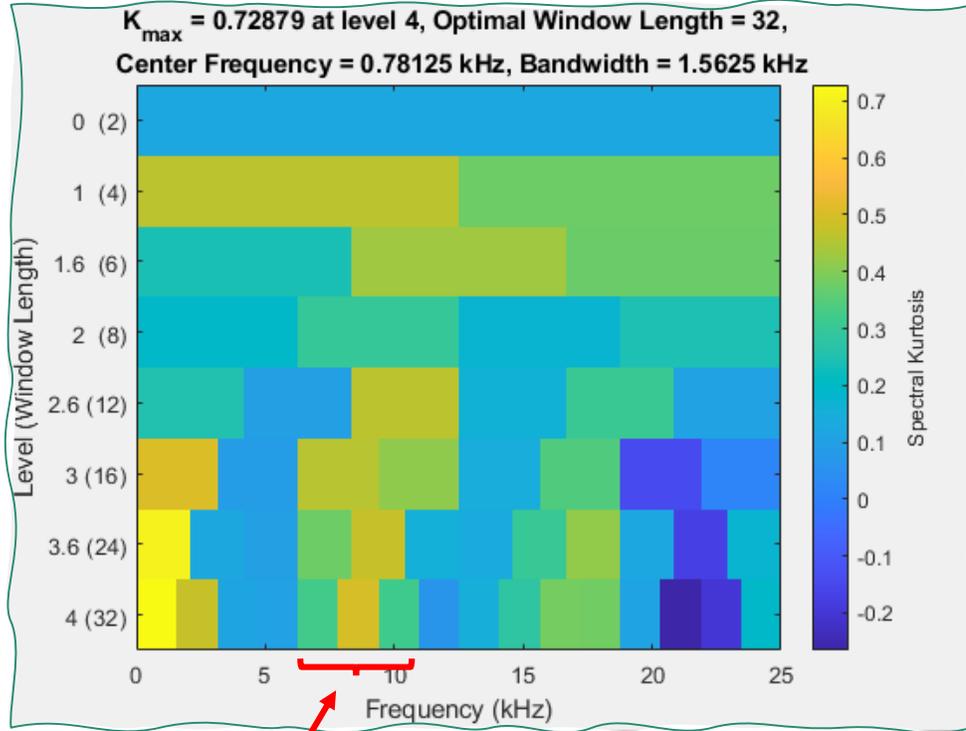


Time Series



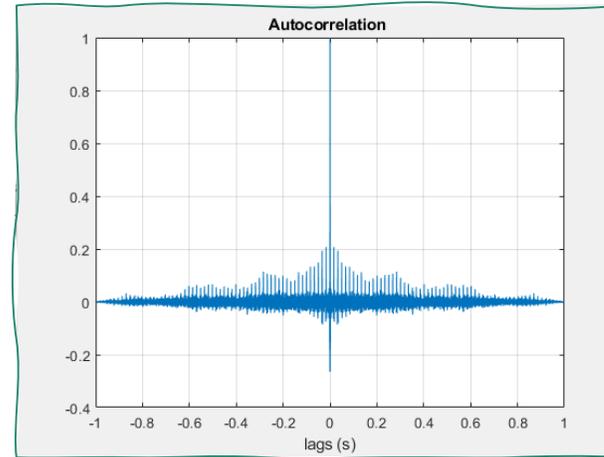
# Ambiguous Kourtogram (second case)

**Extremely low frequency filter bandwidth** (Accelerometer Data from bearing with outer race fault (MAFAULDA). RPM= 61 rev/s)

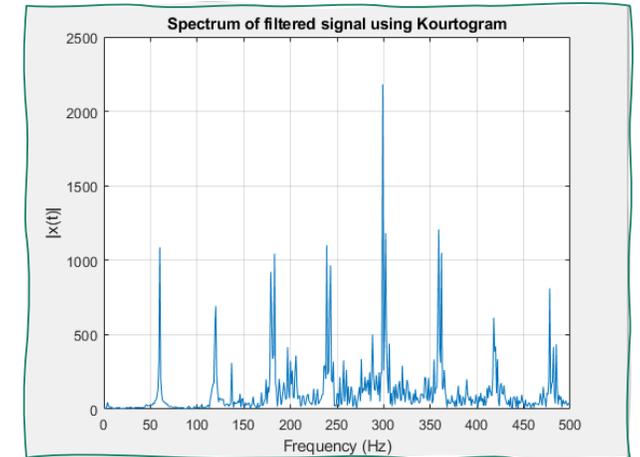


Damage Bandwidth

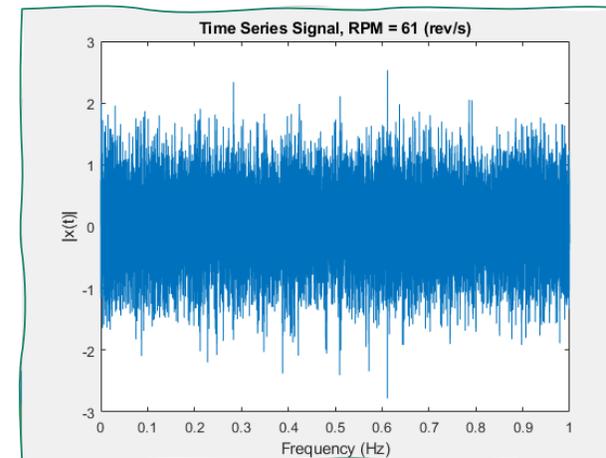
Autocorrelation

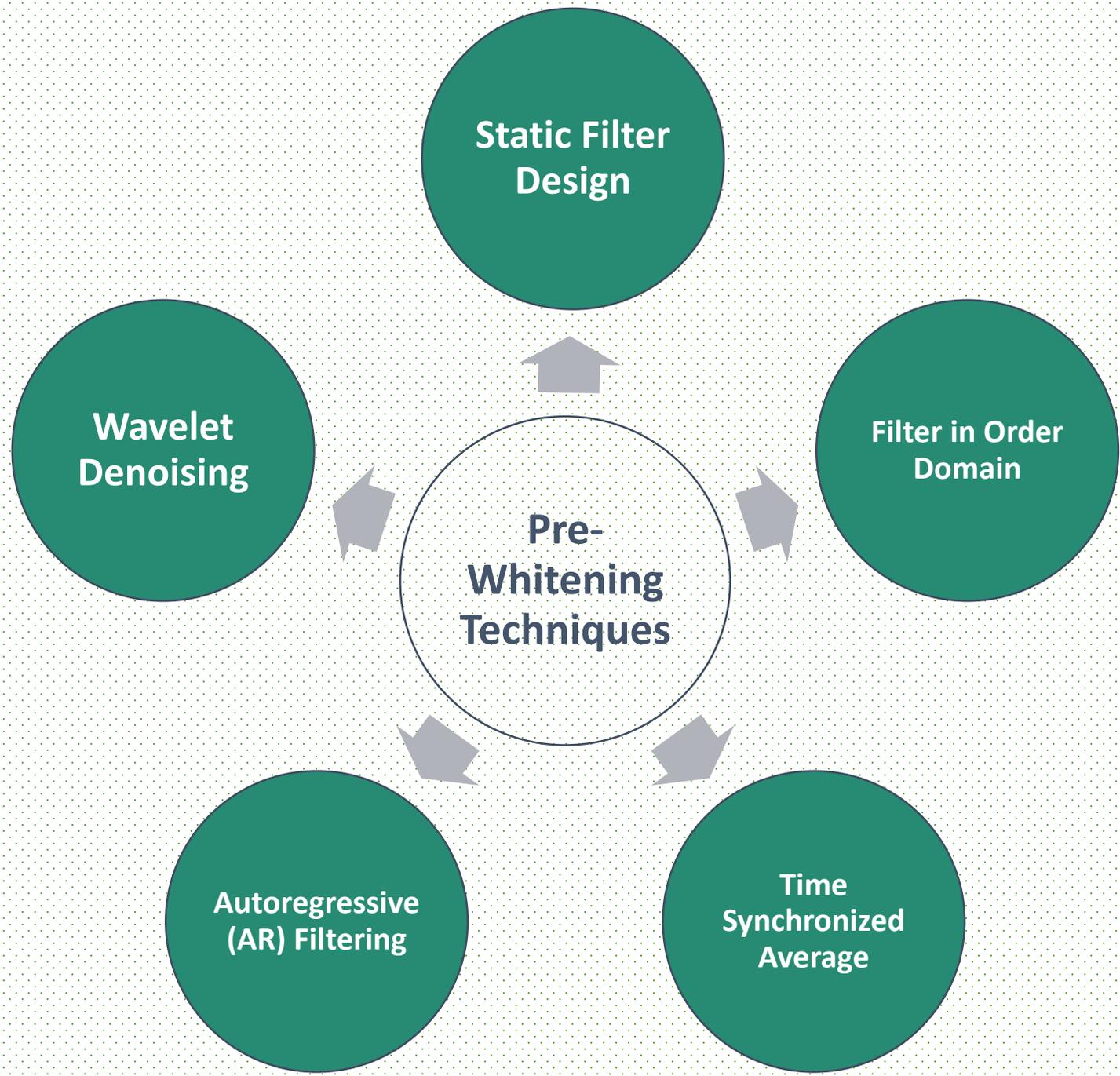
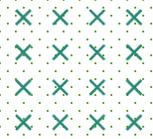


Filtered Spectra using Kourtogram



Time Series





# Static Filter Design

- **Steps:**

- 1- Compare the spectra of the signal in presence and absence of the fault
- 2- Look for a separation between the two signal
- 3- Set a threshold from which separation is meaningful (6-8dB suggested by Randall (1985))

3- Design a filter to extract amplified part of the signal

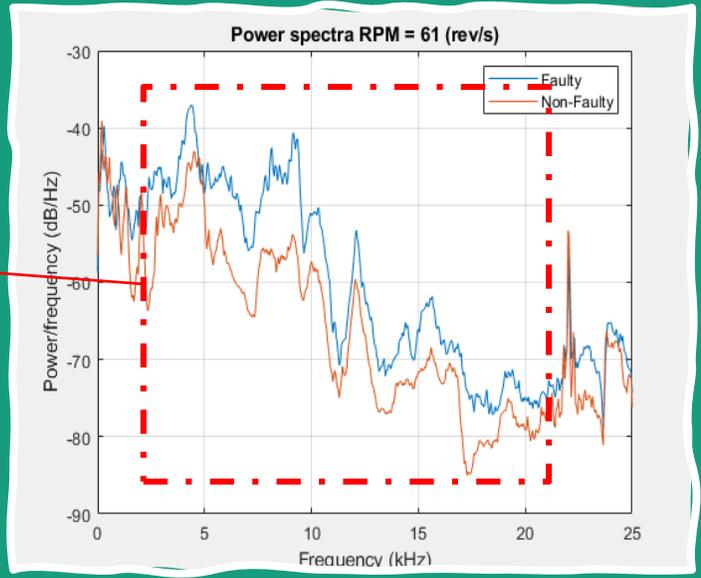
4- calculate Kortugram

**How I have done it (after filtering)**

- **Pros and Cons:**

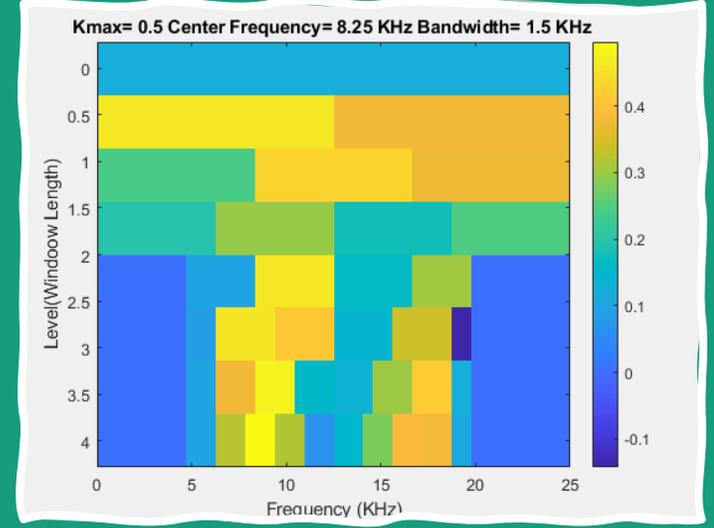
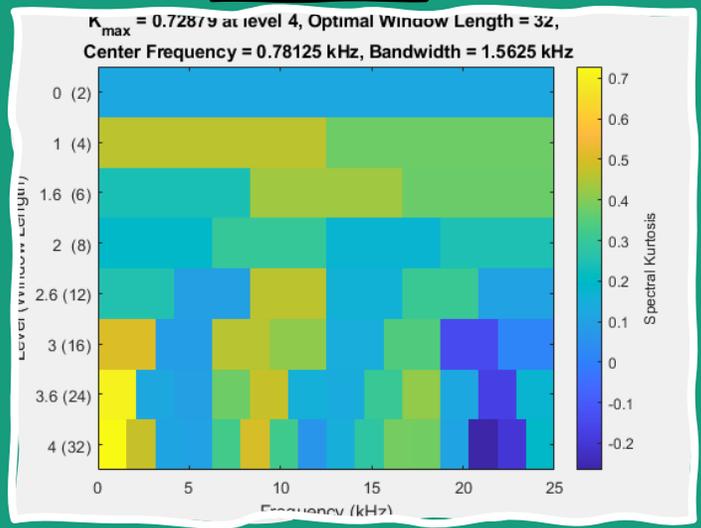
- 1- Simple
- 2- Not accurate
- 3- Prefiltering can cause spurious values
- 4- Slow
- 5- Necessity of base line

6-8dB separation  
(see: Randall (1985))



**Before**

**After**



# Time synchronized average

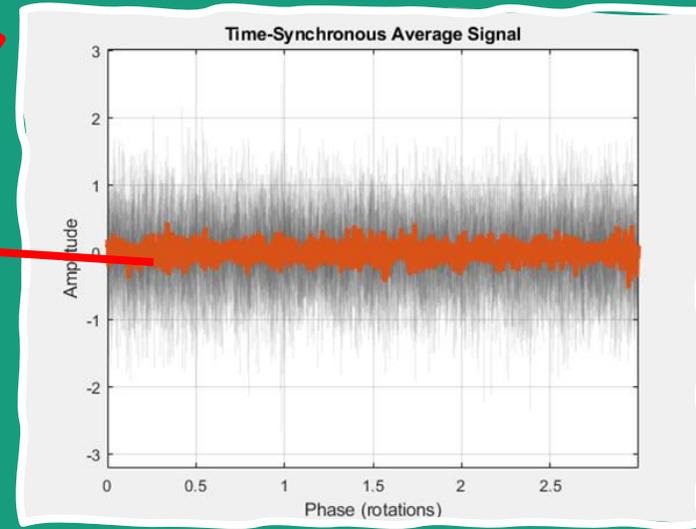
- **Steps:**

- 1- Angular Resampling
- 2- Segmenting signal into the revolutions
- 3- Summing up and averaging out, over revolutions
- 4- Bring back to time domain
- 5- subtract the averaged signal

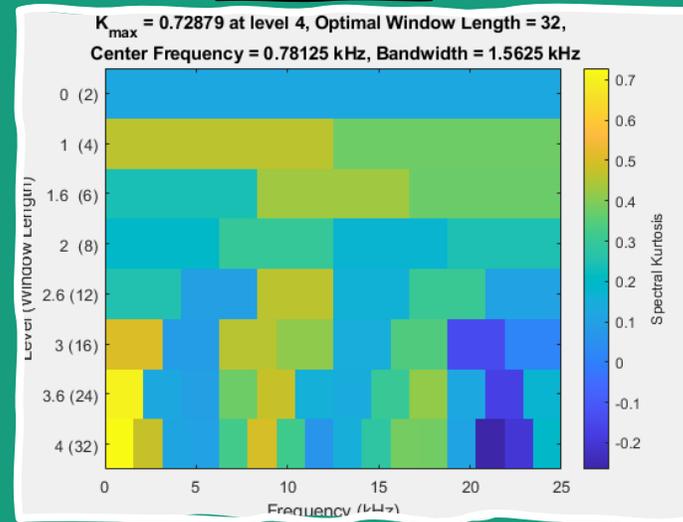
- **Pros and Cons:**

- 1- Complexity
- 2- not stable
- 3- the most known technique
- 4- removes cyclo-stationary signatures such as bearing fault
- 5- prone to noise

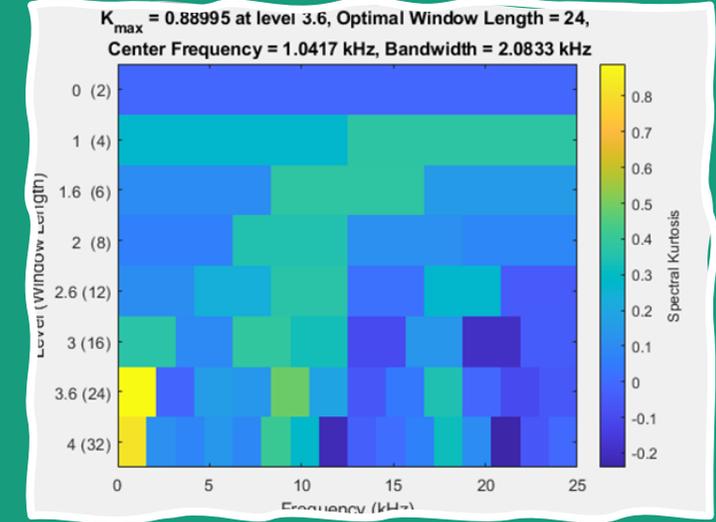
Stationarity Extracted by TSA



Before



After



# Wavelet Denoising

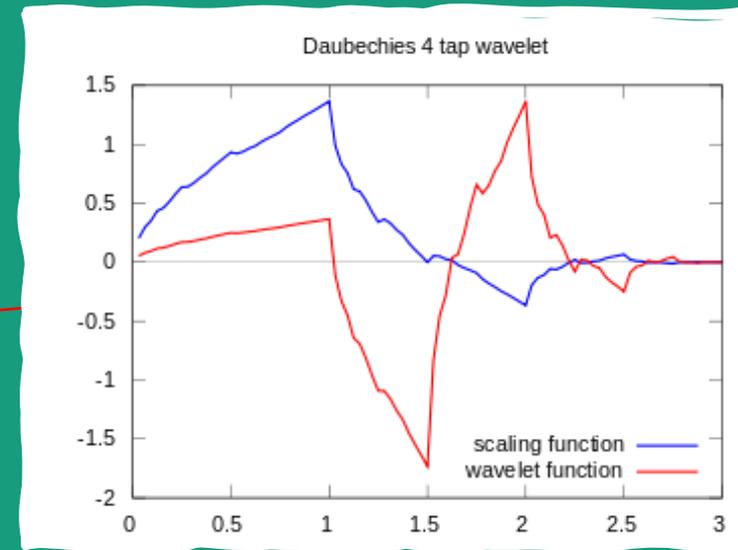
- **Steps:**

- 1- Find the suitable wavelet and wavelet parameters
- 2- Denoise the original signal using wavelet
- 3- Subtract the Denoised signal from original signal.
- 4-The residual of Denoised signal, in fact is a representation of the noise and the nonstationary part of the signal.

- **Pros and Cons:**

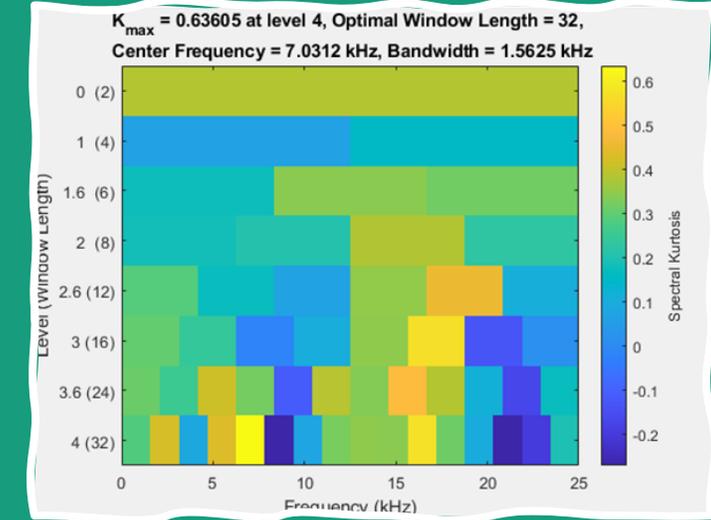
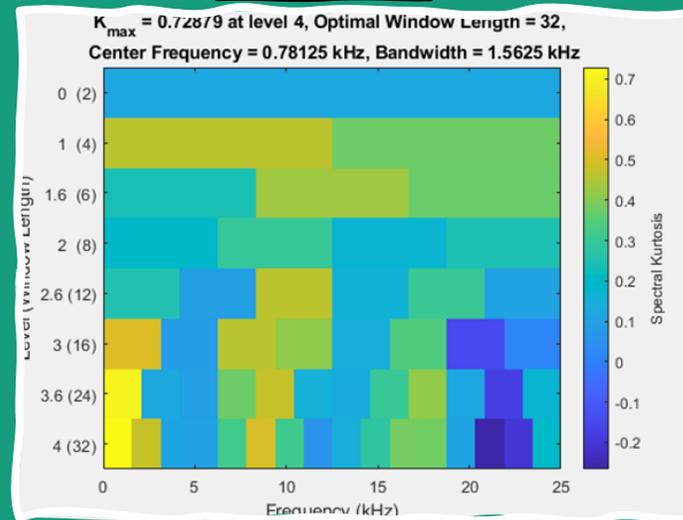
- 1- not suitable for low SNR signals
- 2-Experties Needed
- 3- Can be Atomized

Residual signal from Denoising process using db-4 wavelet for 12 level, BlockJS method



**Before**

**After**



# Filter in order domain

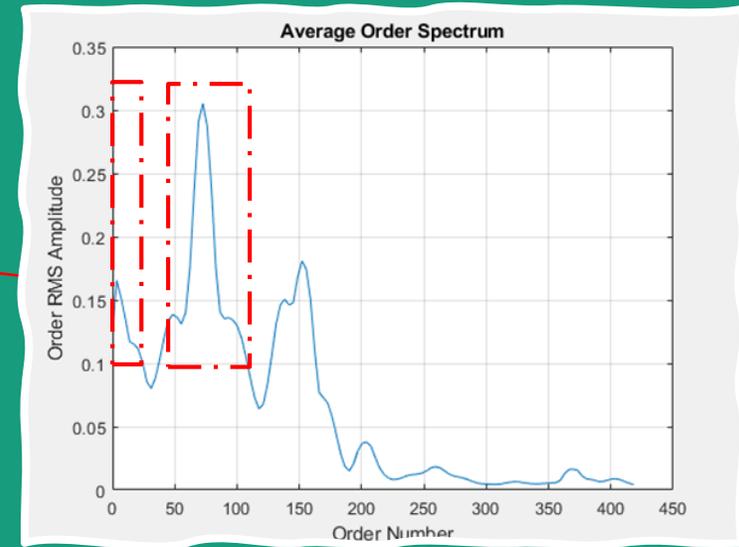
- **Steps:**

- 1- Angular Resampling
- 2- Calculate the order spectra
- 3- set the threshold
- 4- Remove the order picks

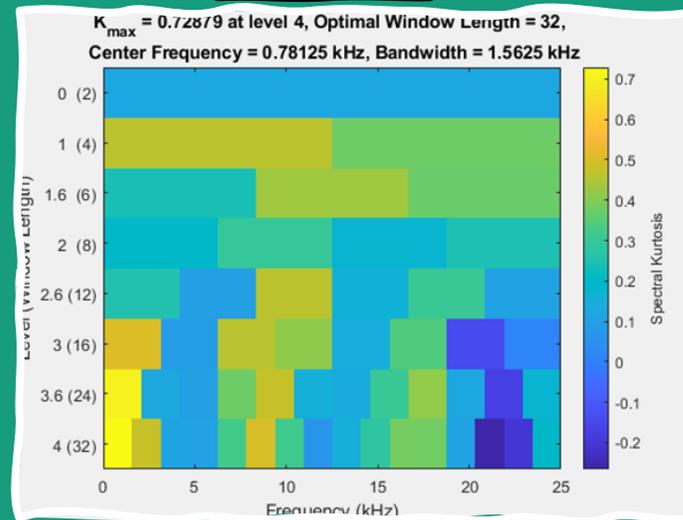
- **Pros and Cons:**

- 1- sensitive to the threshold
- 2- Expertise Needed
- 3- important information might be removed

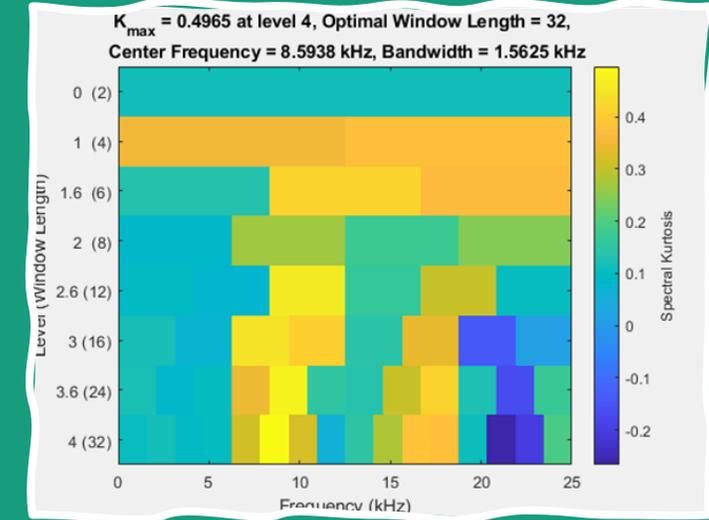
Removed picks



Before



After



# Autoregressive (AR) Model

- **Steps:**

- 1- Find the suitable AR Model
- 2- Optimizing the order
- 3- Generating the model
- 4- Estimating the stationary signal
- 5- Computing the error

- **Pros and Cons:**

- 1- Expertise needed in modeling
- 2- prone noise

Mathematical Representation

$$X_t = \sum_{i=1}^p \varphi_i X_{t-i} + \varepsilon_t$$

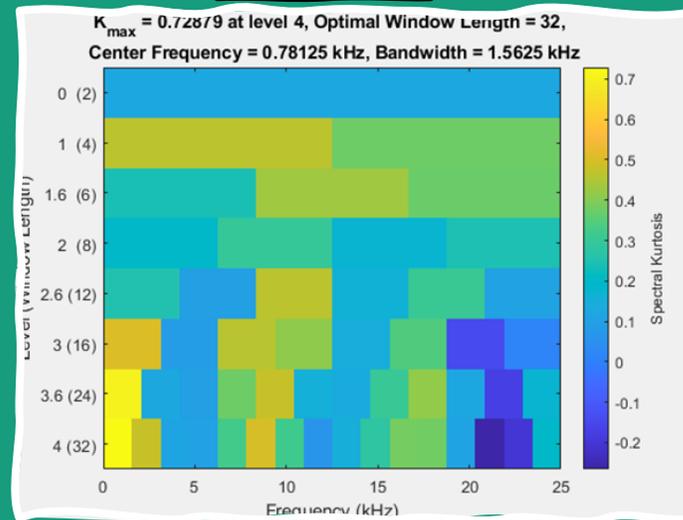
$p$  : Model order (number of lag)

$X_{t-i}$ : Lagged Value

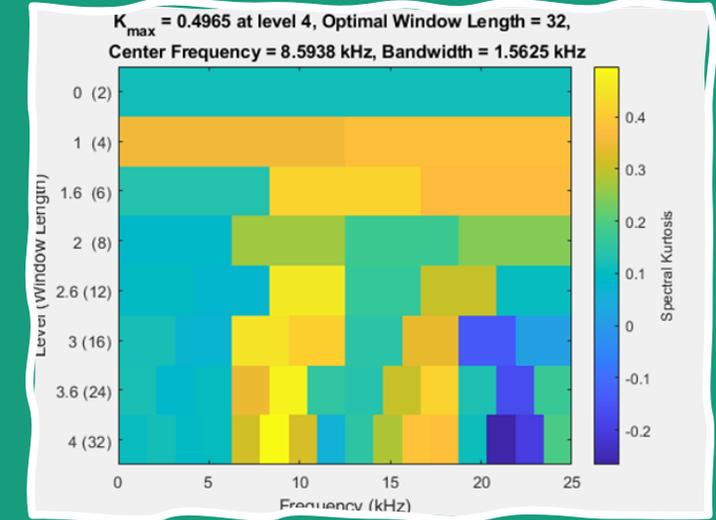
$\varepsilon_t$ : Prediction error (**noise + non stationarity**)

$\varphi$ : Model parameter

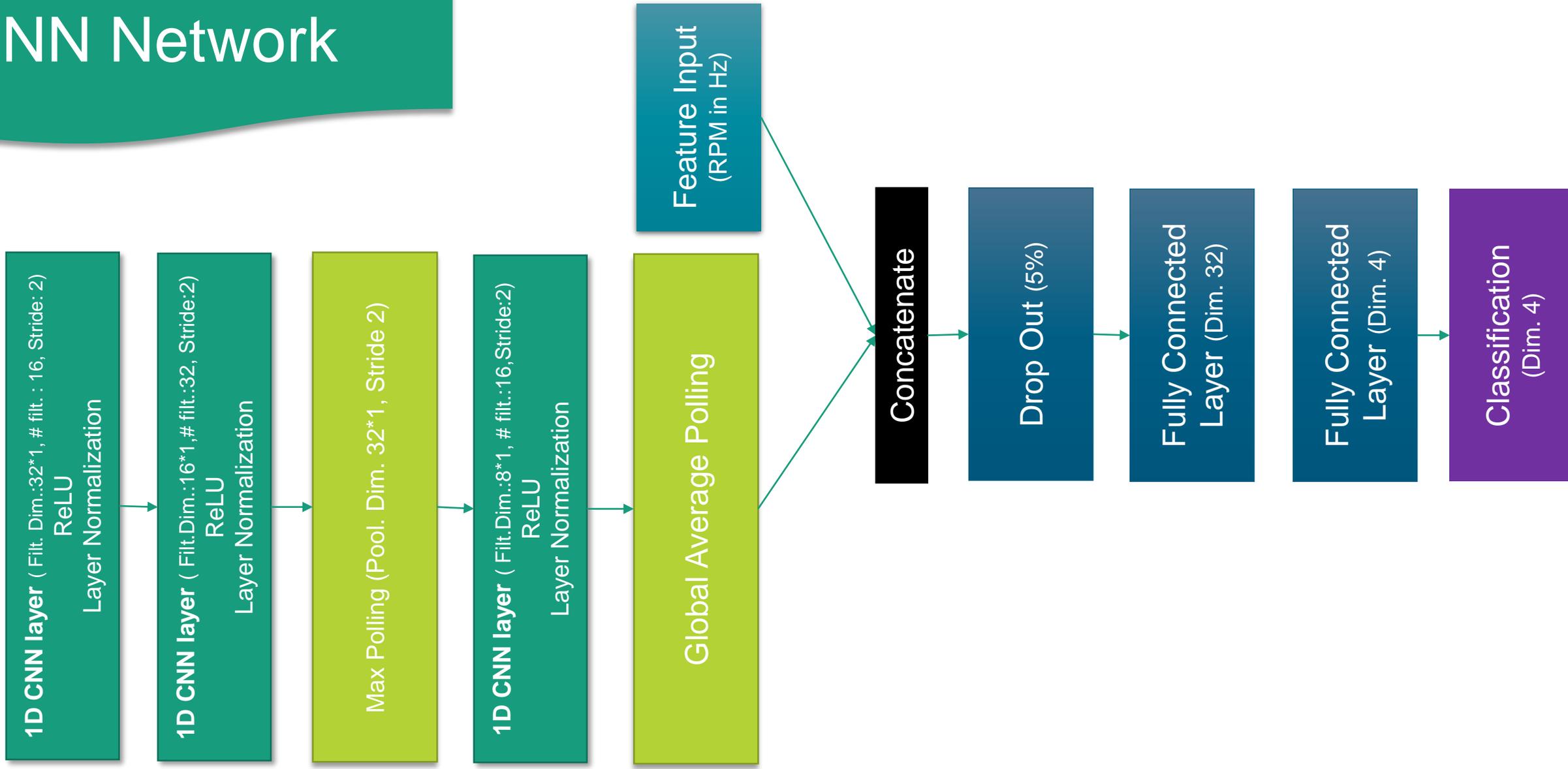
Before



After

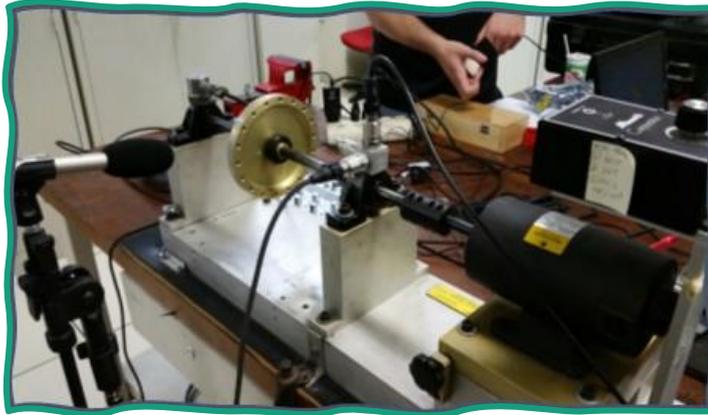


# CNN Network



# MAFAULDA Dataset

- Test bench: SpectraQuest's
- Bearings: two ABVT 8 rolling ball
- RMP range: 700 – 3600rpm
- Sensors:
  - Tachometer
  - Two tridimensional accelerometer
  - Microphone
- Sampling Frequency: 50 KHz
- Measurement duration: 5s
- Outer/Inner/cage fault scenario



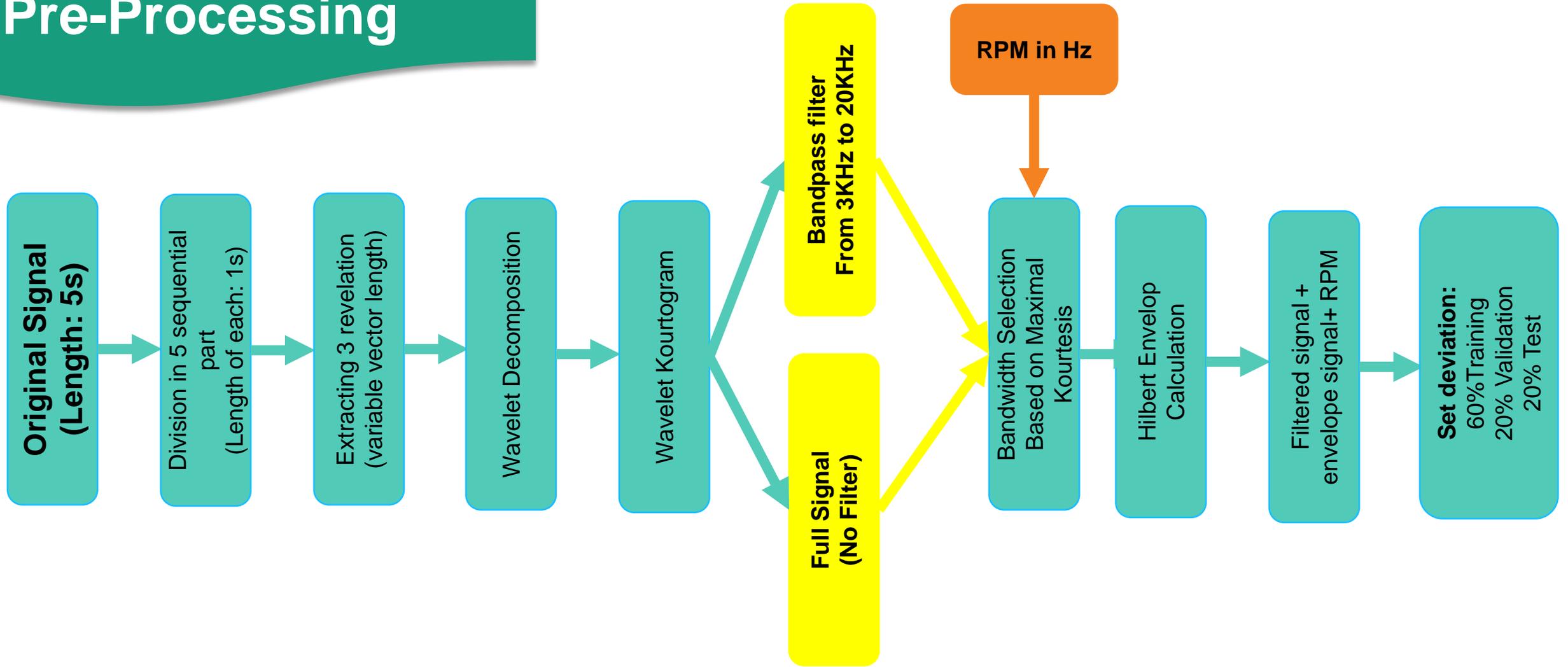
SpectraQuest's Machinery Fault Simulator (MFS)

Weight (g)	Measurements
6	49
10	48
15	48
20	49
25	47
30	47
35	45
<b>Total</b>	<b>333</b>

Imbalance Load details

Tachometer '	'Underhang Bearing Accelerometer Axial'	'Underhang Bearing Accelerometer Radiale '	'Underhang Bearing Accelerometer Tangential '	'Overhang Bearing Accelerometer Axial'	'Overhang Bearing Accelerometer Radiale '	'Overhang Bearing Accelerometer Tangential '	'microphone'
-0.56205	0.49038	-0.047927	0.048199	0.16672	0.034679	0.78593	0.10149
-0.56537	-1.5267	-0.11375	0.017286	0.16381	0.035058	0.73414	-0.0027973
-0.56313	0.971	0.23181	0.063505	0.1694	0.037785	0.80038	-0.01078
-0.5625	-1.5916	-0.44822	0.010677	0.16446	0.036373	0.68713	0.14934
-0.56661	1.0665	0.35208	0.10047	0.1659	0.03672	0.74835	-0.13203
-0.55848	-0.97374	-0.45289	0.0046073	0.16389	0.032985	0.68246	0.13768
-0.56572	0.3661	0.35023	0.030267	0.1682	0.03722	0.72353	-0.15036
-0.56366	-0.28867	-0.10572	-0.031329	0.16652	0.034628	0.67451	0.13157
-0.55979	-0.90231	0.22798	-0.00416	0.17068	0.041341	0.64372	-0.096815
-0.56708	0.66282	0.15316	-0.001411	0.17285	0.042277	0.66084	0.075668
-0.55558	-1.7188	-0.14494	0.04	0.17252	0.045812	0.58502	0.048239

# Data Handling & Pre-Processing

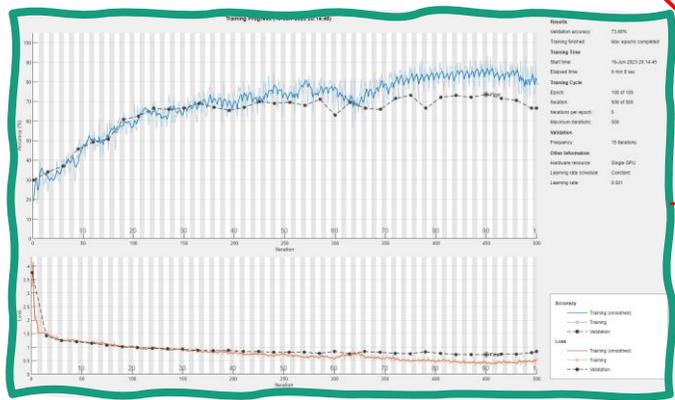


# Case Study

## No Pre-Whitening

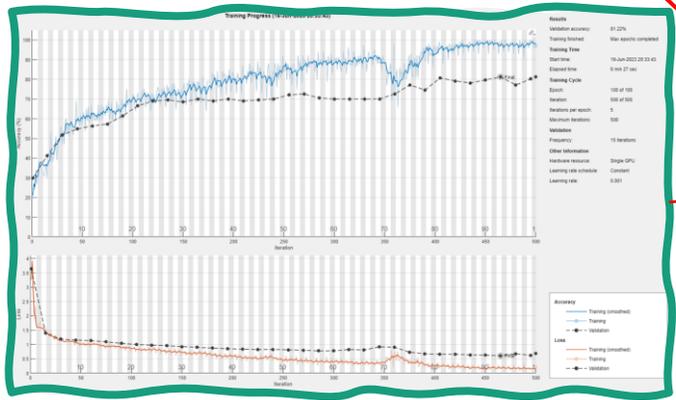
## Bandpass Filter

Learning Curve



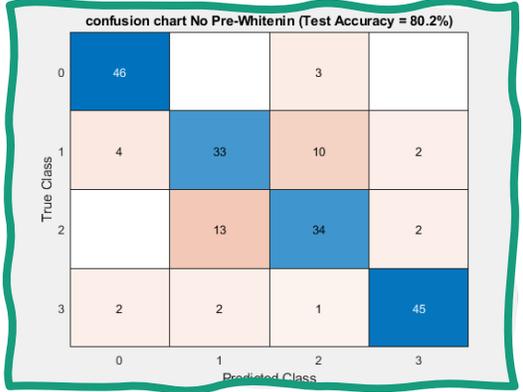
Test Accuracy = 80.2%

Learning Curve

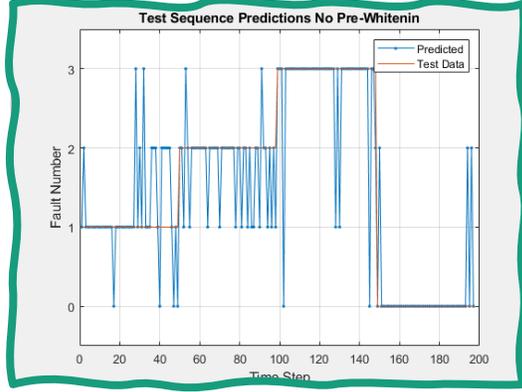


Test Accuracy = 84.3%

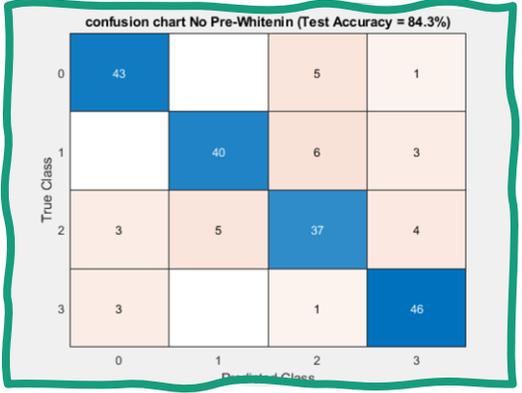
Confusion Chart



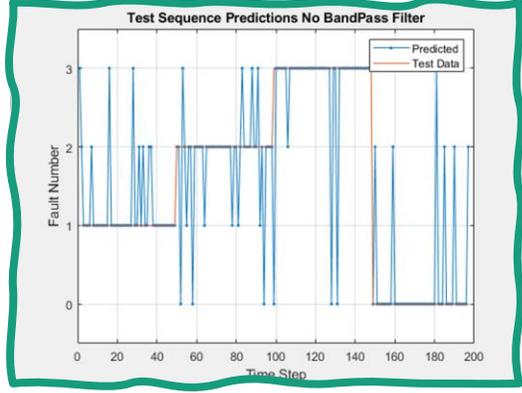
Test Sequence Predictions



Confusion Chart



Test Sequence Predictions



# Thank you!

DO YOU HAVE ANY QUESTIONS?



The Autor Gratefully acknowledge the European commission for its support of the maria sklodowska curie program through the ETN MOIRA project (GA 955681)