

The Seventeenth International Conference on  
Condition Monitoring and Asset Management

**CM2021**

The future of condition monitoring

Monday 14 to Friday 18 June 2021



This is now a  
virtual event

# Optimal Feature Performance Analysis in IESFOgram for Bearing Diagnostics

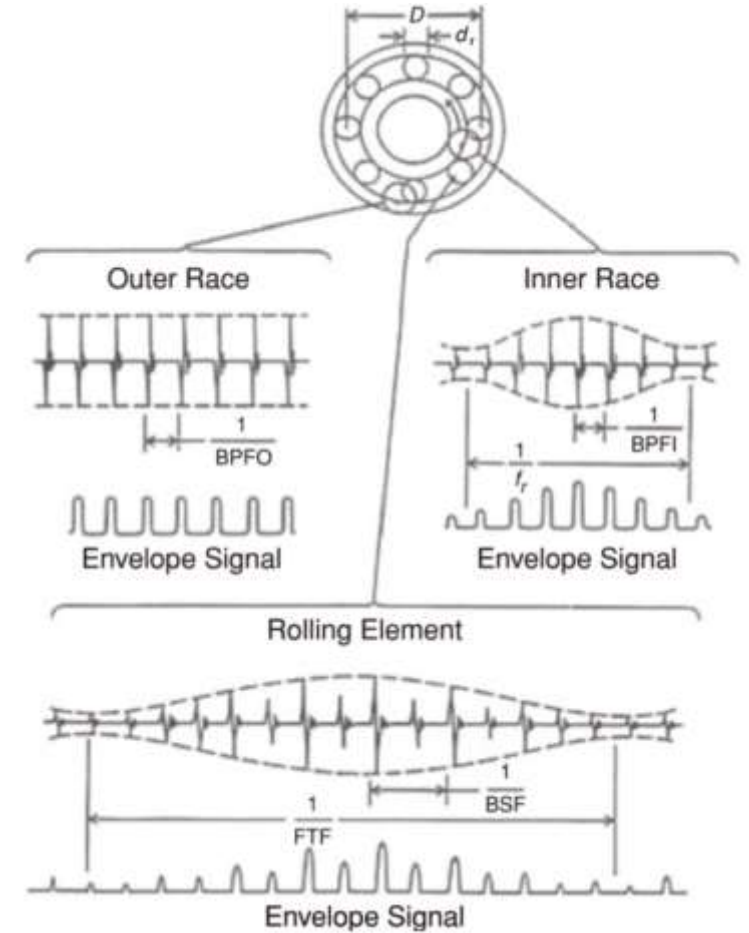
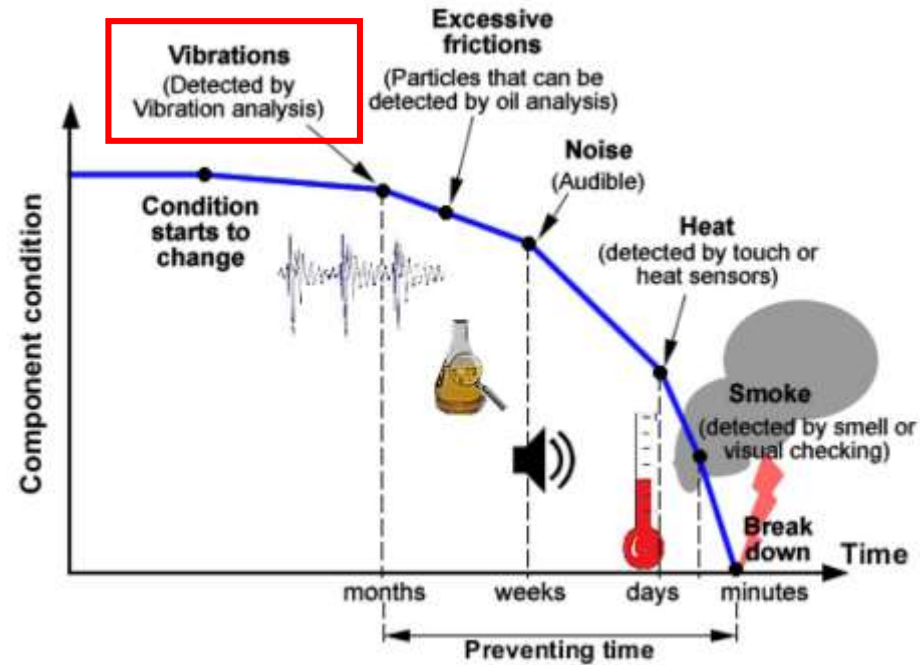
A. Mauricio<sup>1,2</sup>, S. Schmidt<sup>3</sup>, K. Gryllias<sup>1,2</sup>

<sup>1</sup>Department of Mechanical Engineering, KU Leuven

<sup>2</sup>DMMS-M: Dynamics of Mechanical and Mechatronic Systems, Flanders Make

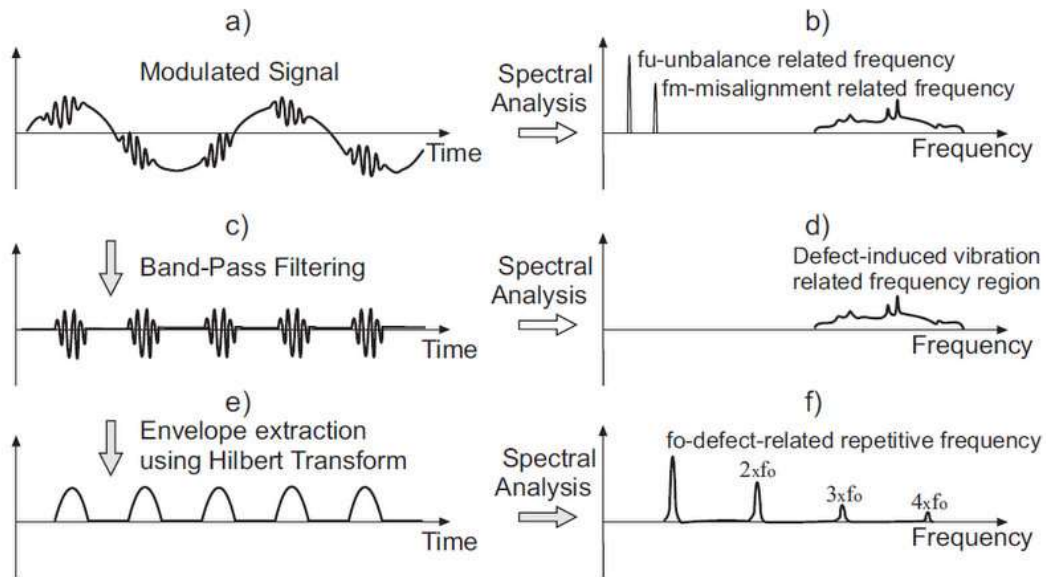
<sup>3</sup>Centre for Asset Integrity Management, Department of Mechanical and Aeronautical Engineering, University of Pretoria, Pretoria, South Africa

# Condition monitoring

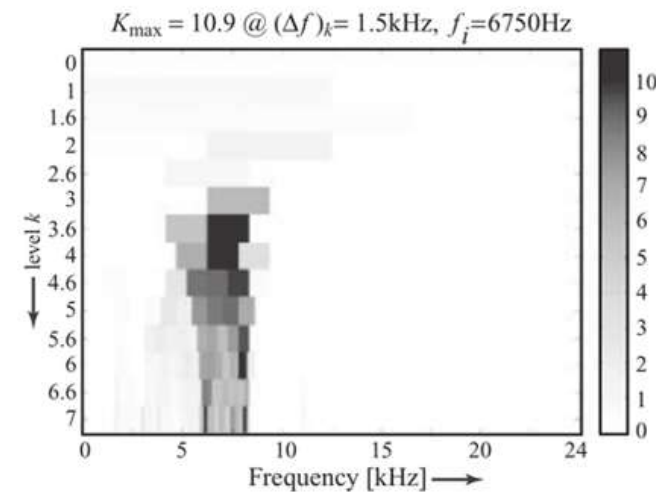


# Bearing damage detection

## Filtered SES



## Fast Kurtogram



## Blind Features:

Fast Kurtogram

Autogram

Protrugram

Infogram

...

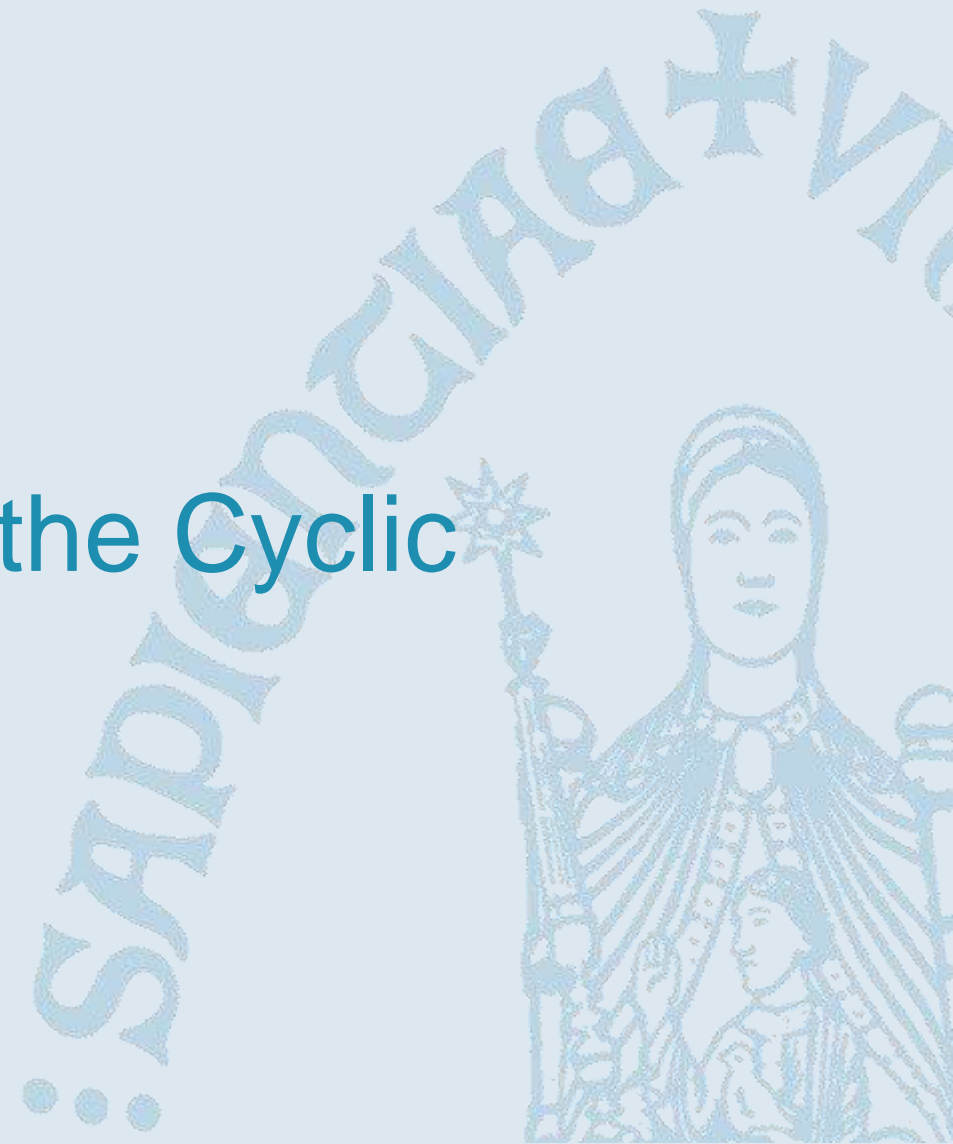
## Frequency Targeted:

Cyclogram

ICS2gram

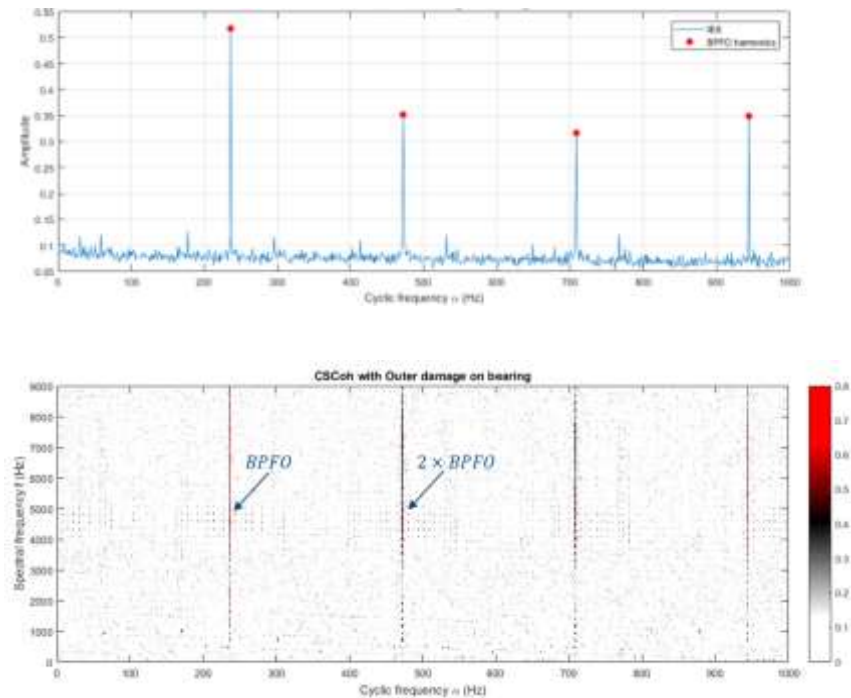
IFBIgram

# Blind and Targeted features on the Cyclic Spectral Coherence

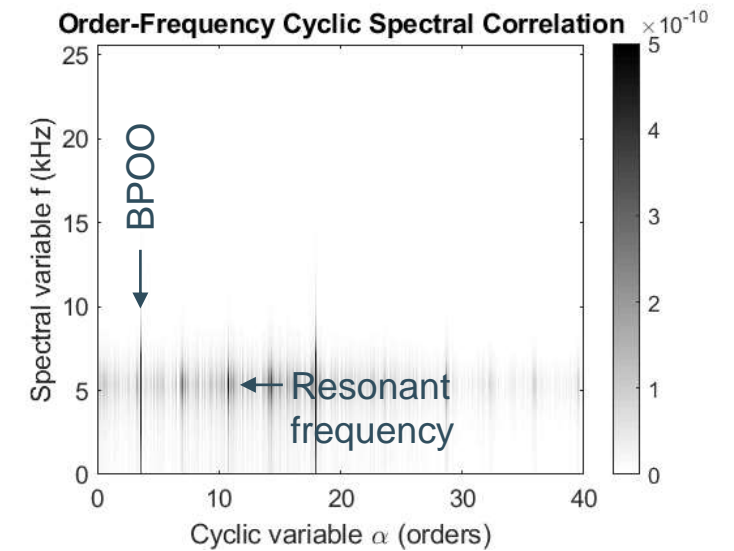
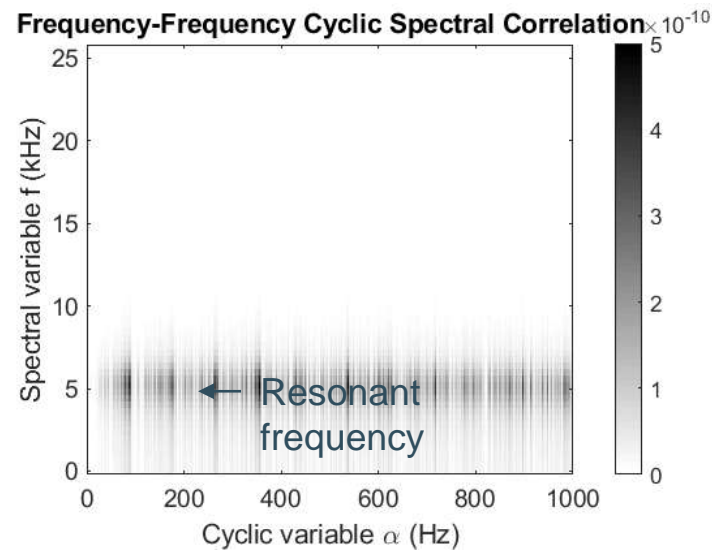


# Steady and varying speed for diagnostics

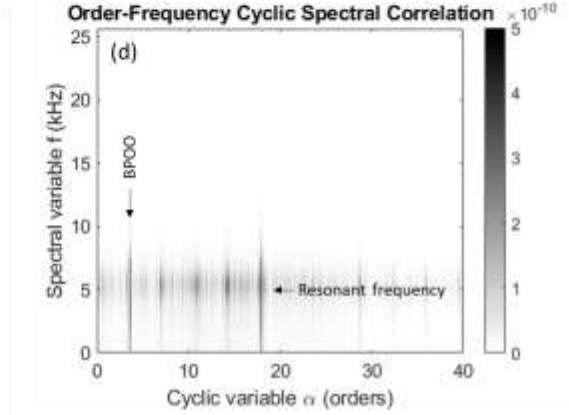
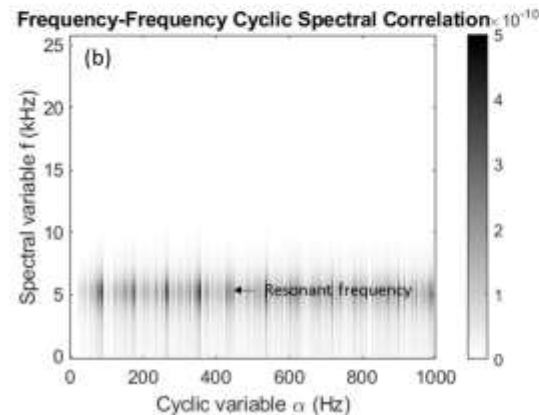
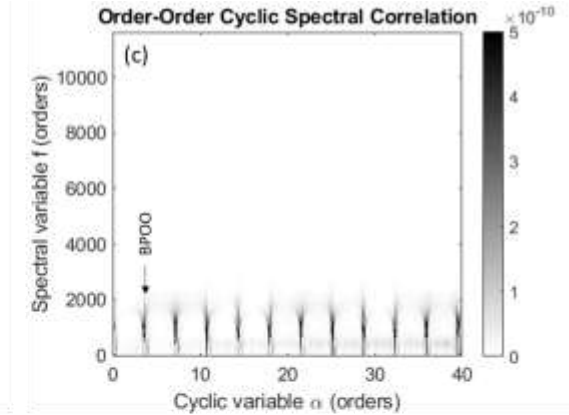
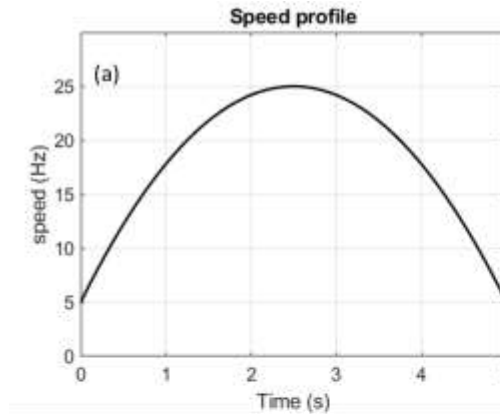
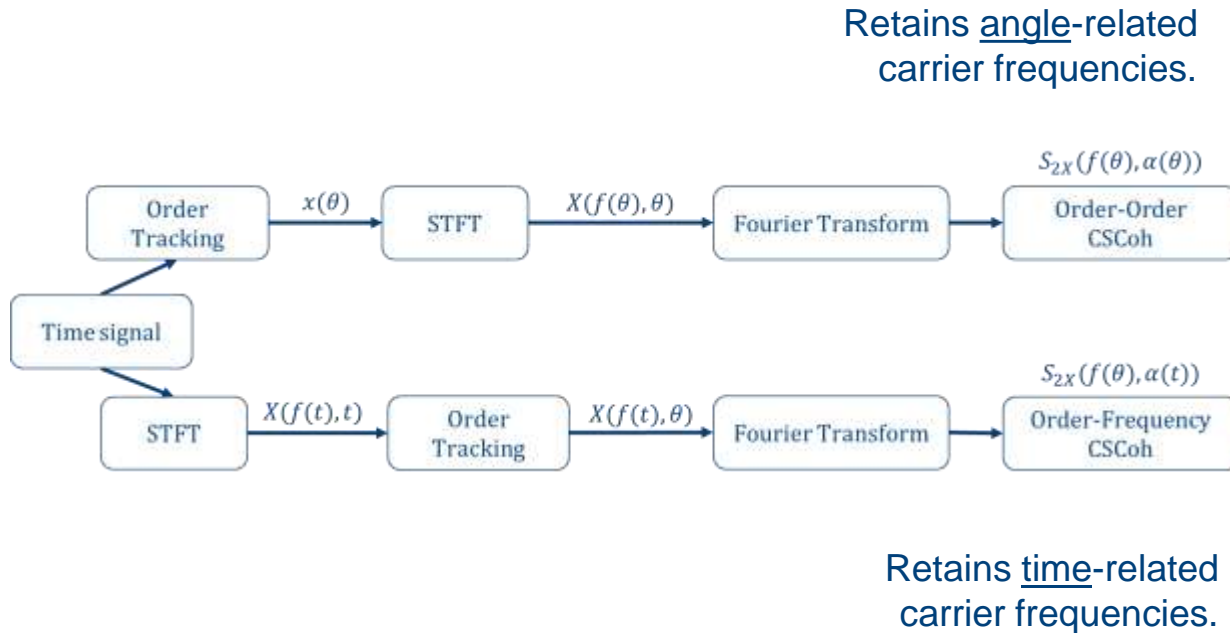
Steady speed



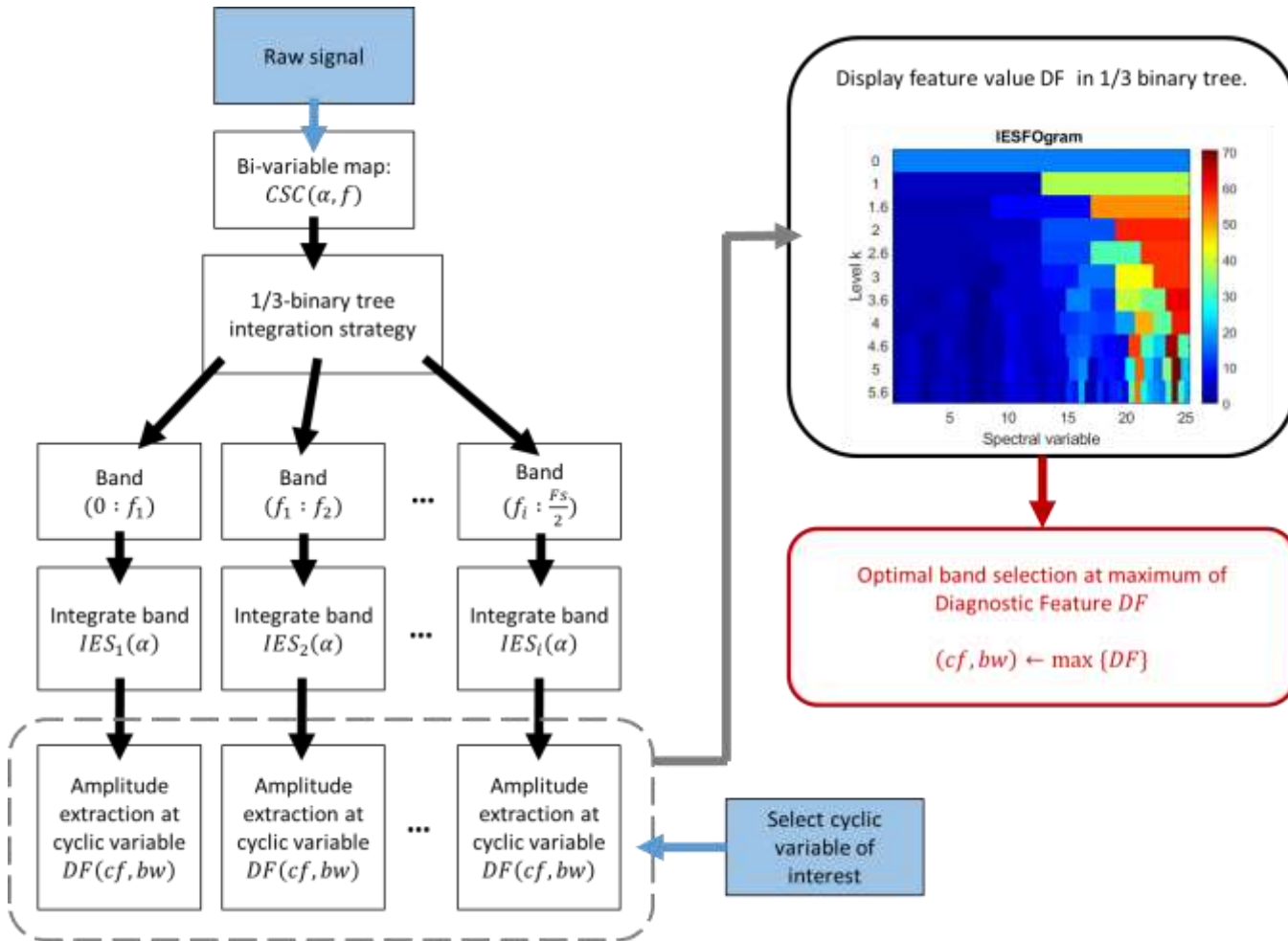
Varying speed



# Steady and varying speed for diagnostics



# Methodology



## Blind features:

Spectral Kurtosis,  
Spectral Negentropy  
Gini Index  
Spectral Flatness

$$K = \frac{\frac{1}{N} \sum_{i=1}^N (x_i - \bar{x})^4}{(\frac{1}{N} \sum_{i=1}^N (x_i - \bar{x})^2)^2} - 3$$

$$Neg = \frac{1}{N} \sum_{i=1}^N \left( \frac{x_i^2}{\frac{1}{N} \sum_{i=1}^N x_i^2} \ln \frac{x_i^2}{\frac{1}{N} \sum_{i=1}^N x_i^2} \right)$$

$$G = 1 - 2 \frac{1}{N} \sum_{i=1}^N \frac{x(i)}{\|x\|} \left( \frac{N-i+\frac{1}{2}}{N} \right)$$

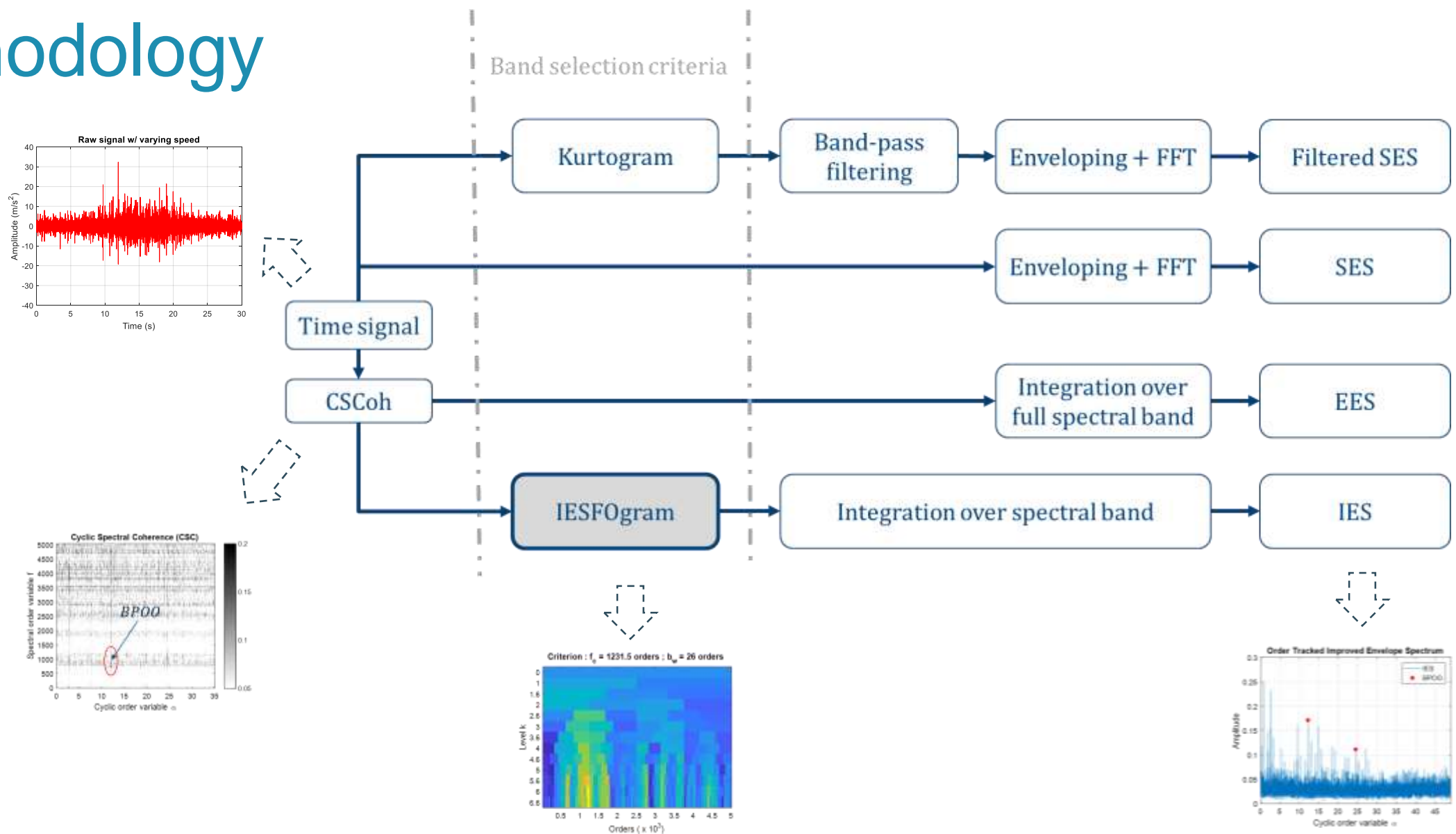
$$SF = \frac{\exp(1/N \sum_{i=1}^{N-1} \ln X(i))}{1/N \sum_{i=1}^{N-1} X(i)}$$

## Bearing Frequency-Targeted features:

$\alpha_{fault} = \text{BPFI, BPFO, BSF or FTF}$

$$DF(cf, bw) = \sum_{k=1}^N \frac{IES_{cf,bw}(k \times \alpha_{fault})}{\frac{1}{2f_b} \left[ \int_{kf_{fault}-f_b}^{kf_{fault}+f_b} IES_{cf,bw}(\alpha) d\alpha - IES_{cf,bw}(k \times \alpha_{fault}) \right]}$$

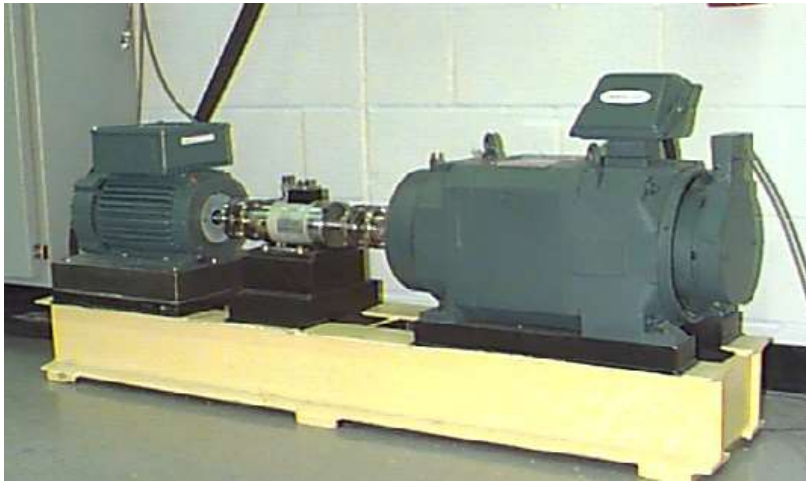
# Methodology



# Experimental Setup



# Case Western Reserve University (open dataset)



The test stand consists of a 2 hp motor (left), a torque transducer/encoder (center), a dynamometer (right), and control electronics (not shown). The test bearings support the motor shaft. Single point faults were introduced to the test bearings using electro-discharge machining with fault diameters of 7 mils, 14 mils, 21 mils (1 mil=0.001 inches)

## Bearing Information

**Drive end bearing:** 6205-2RS JEM SKF, deep groove ball bearing

Size: (inches)

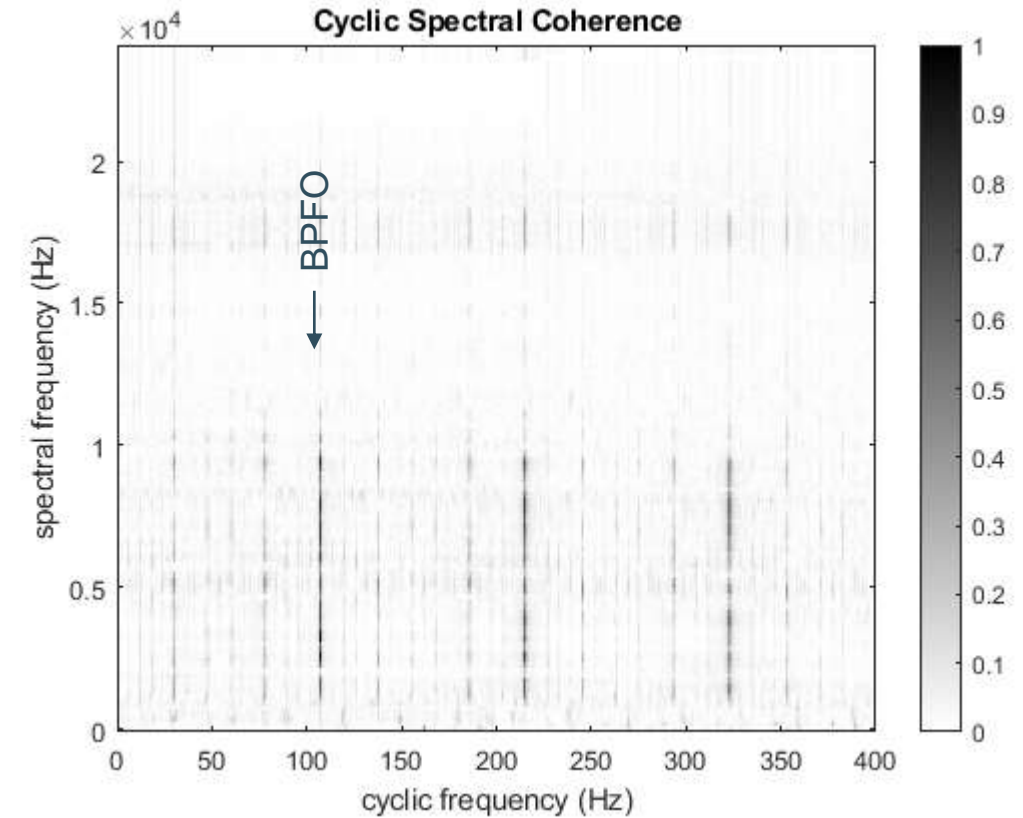
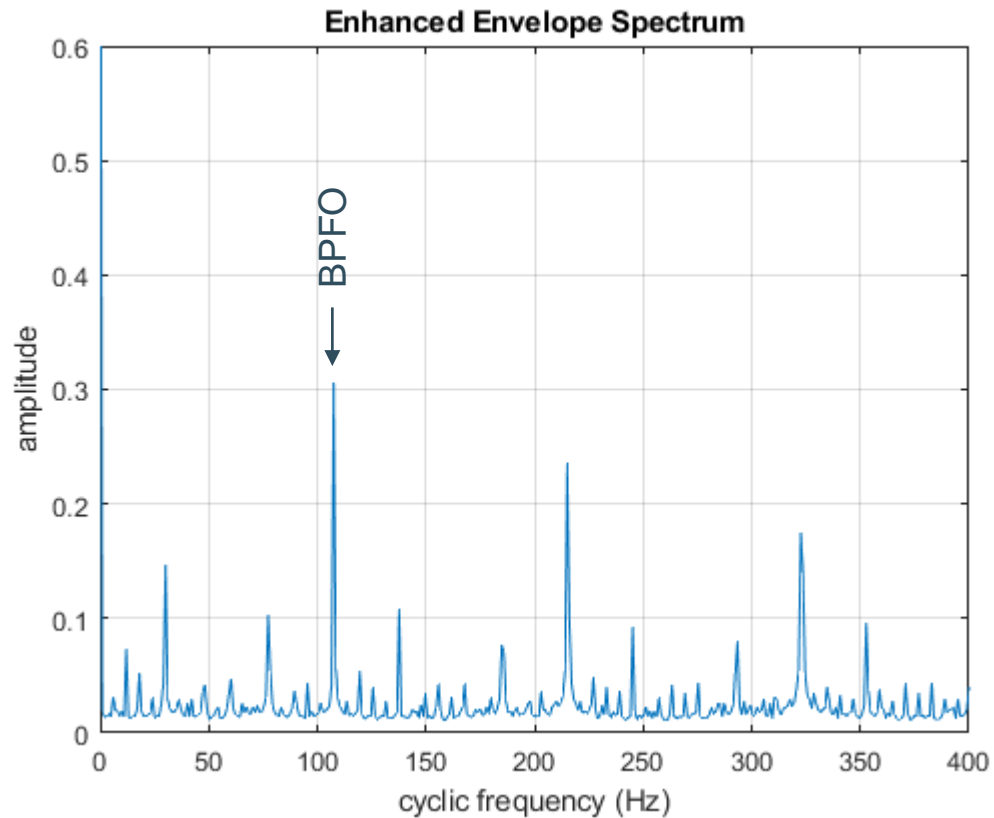
Inside Diameter	Outside Diameter	Thickness	Ball Diameter	Pitch Diameter
0.9843	2.0472	0.5906	0.3126	1.537

Defect frequencies: (multiple of running speed in Hz)

Inner Ring	Outer Ring	Cage Train	Rolling Element
5.4152	3.5848	0.39828	4.7135

**Fan end bearing:** 6203-2RS JEM SKF, deep groove ball bearing

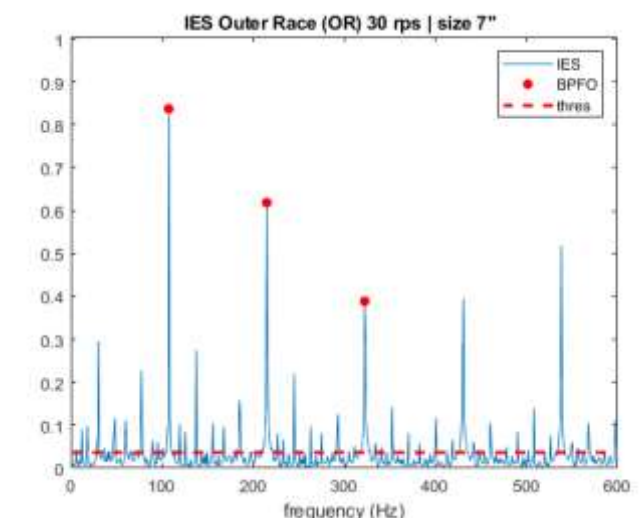
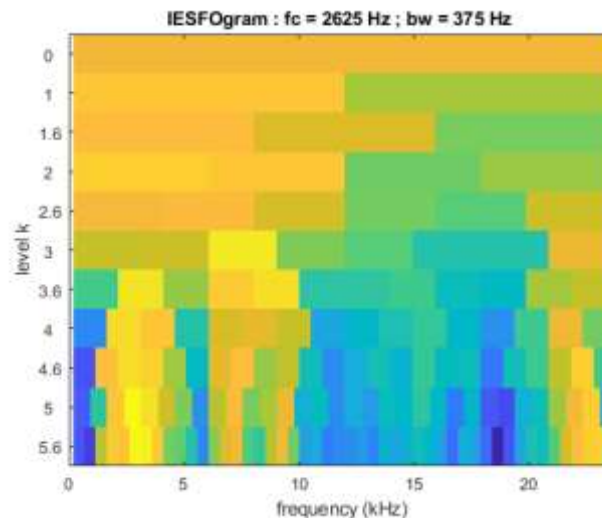
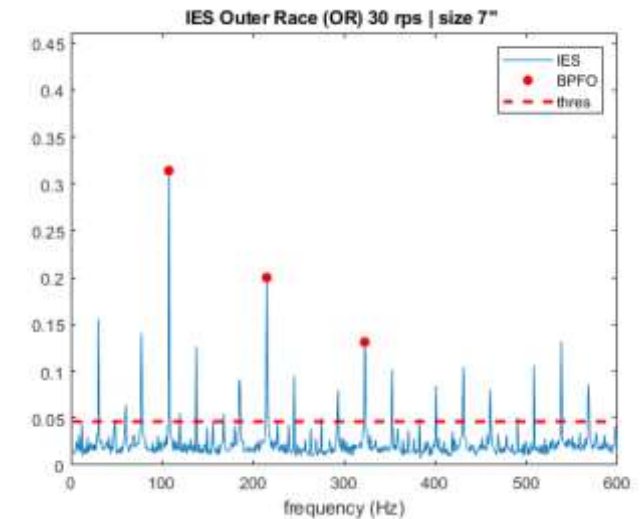
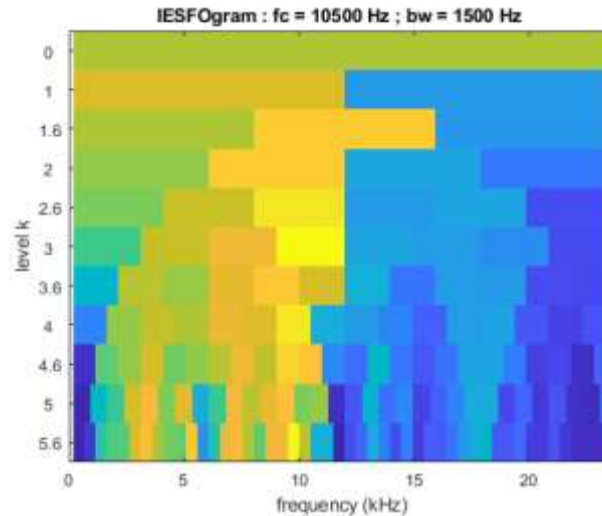
# Outer race damage (7")



# Targeted at BPFO and Gini Index (Blind)

From all features, the targeted feature at the damage related frequency (BPFO) shows a selection of a rich band carrier resulting in clear diagnosis.

The Gini Index also shows good performance on selecting a rich band throughout the dataset.

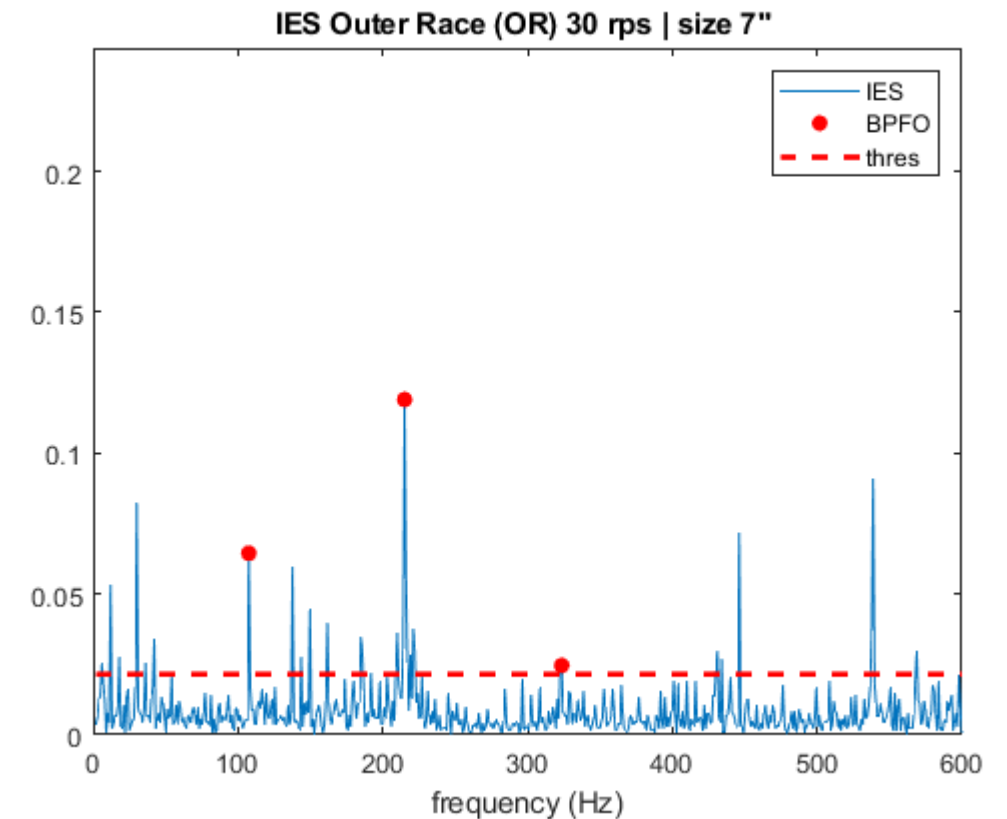
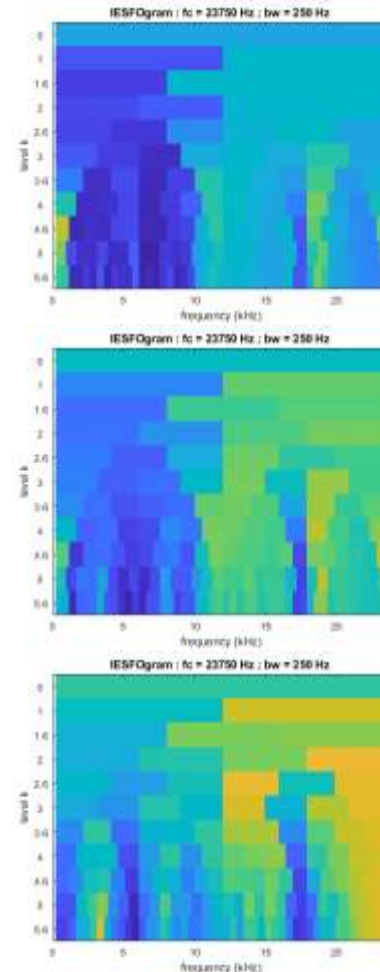


# Remaining blind features

Other blind features select (converge) to small band at the Nyquist.

Envelope spectra at Nyquist is seen often to be sparse.

Kurtosis  
Spectral Negentropy  
Spectral Flatness



# DMMS-M Drivetrain setup



Self-aligning bearing SKF 2206

Damage with width approx. 5.5mm

Made with dremel



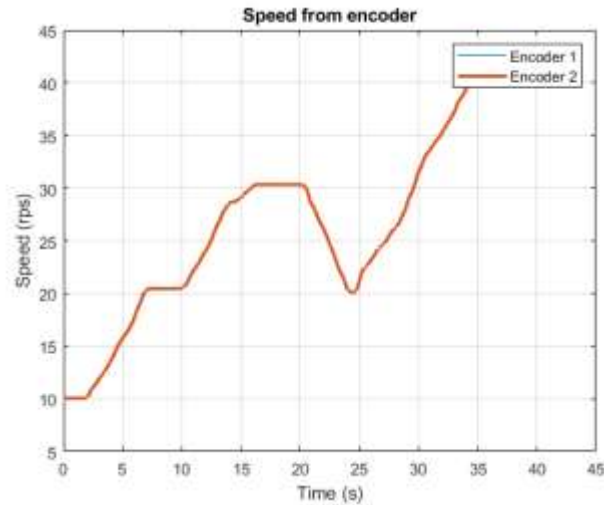
*Characteristic damage frequency:*

*Ball Pass Frequency of Outer race*

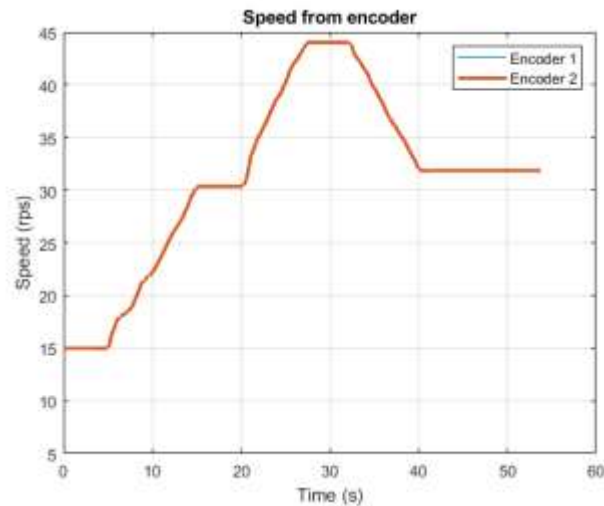
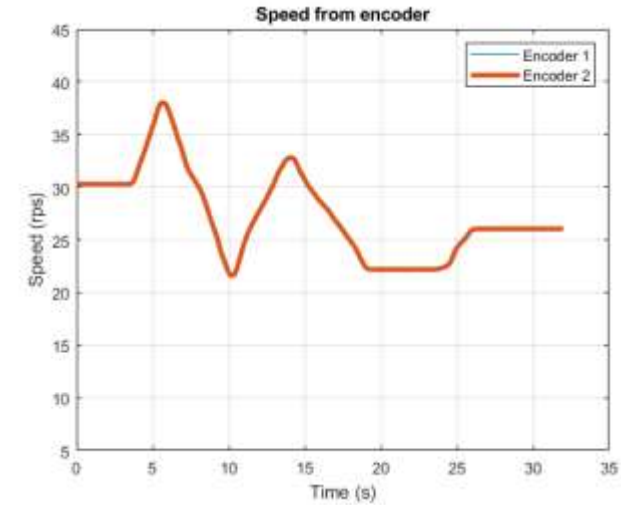
$$BPFO = f_{shaft} * BPOO$$
$$BPOO = 4.724$$

One disc mounted at the middle distance between the bearings (bearing-150mm-disc-150mm-bearing).  
One 3D accelerometer PCB 256A15 mounted on the housing (X-radial horizontal, Y-radial vertical, Z - axial)  
One microphone PCB 378B20 distanced 300mm from housing

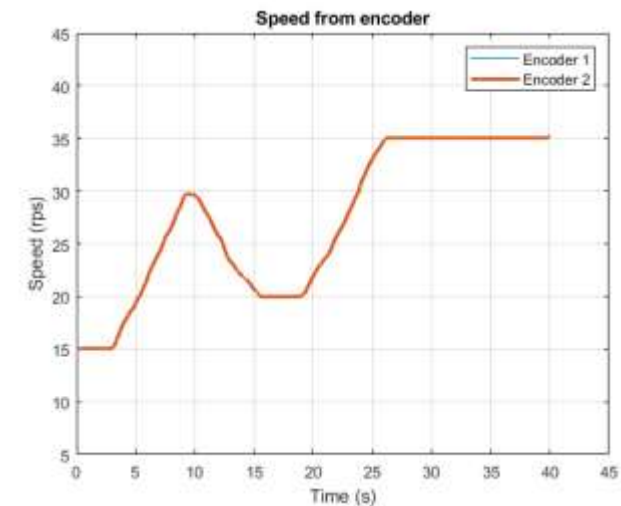
# Varying speed measured



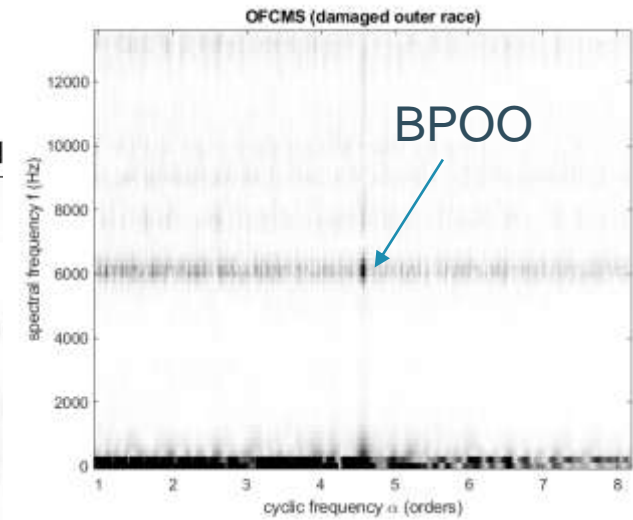
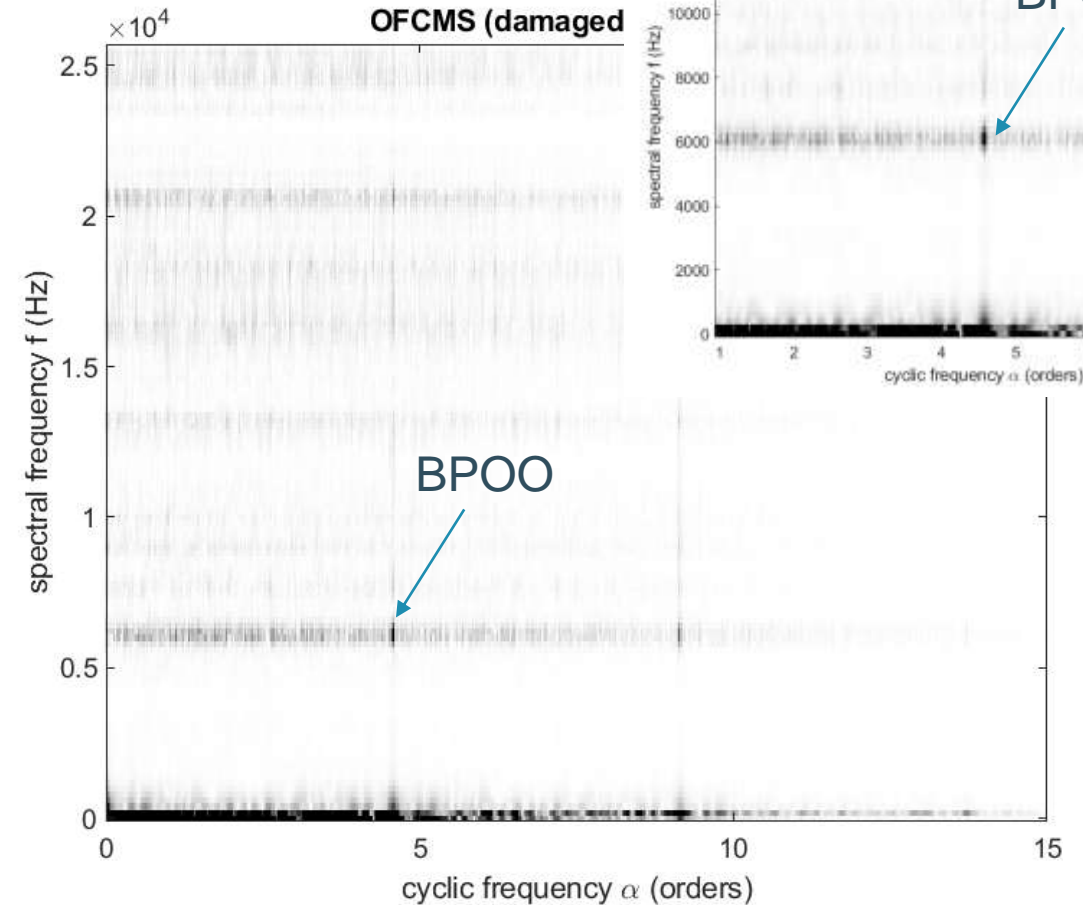
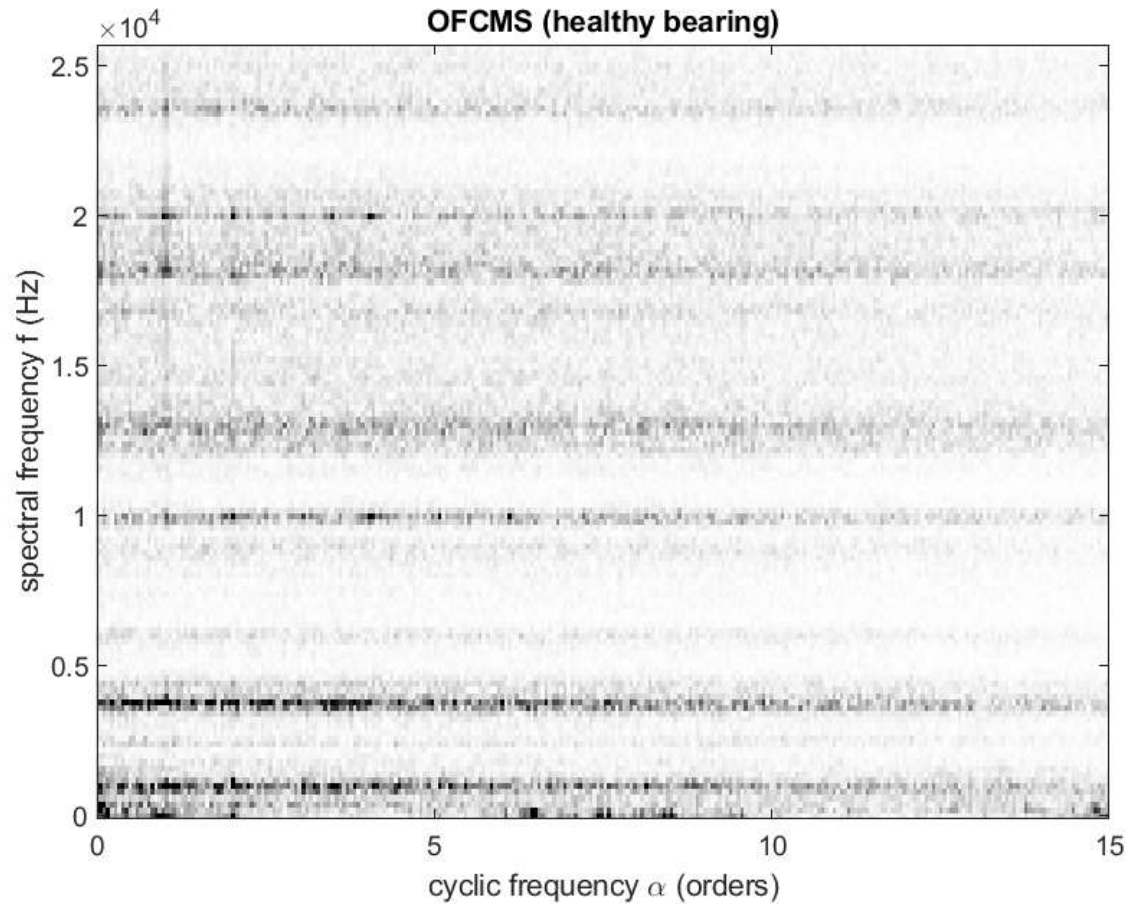
Damaged signals



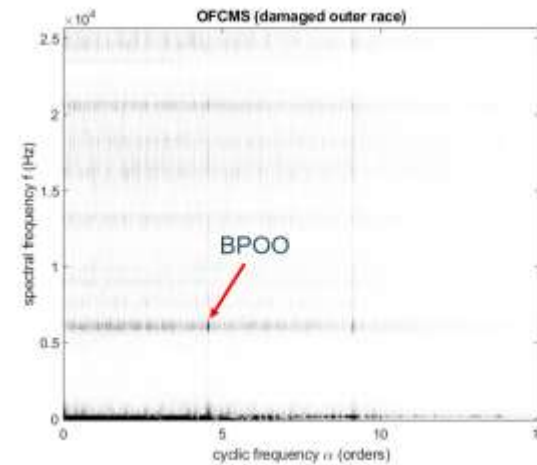
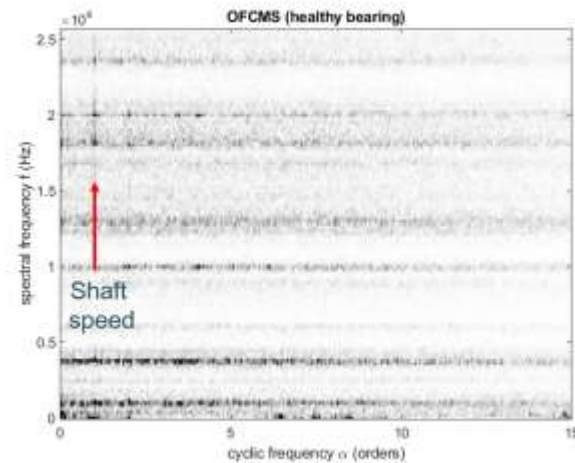
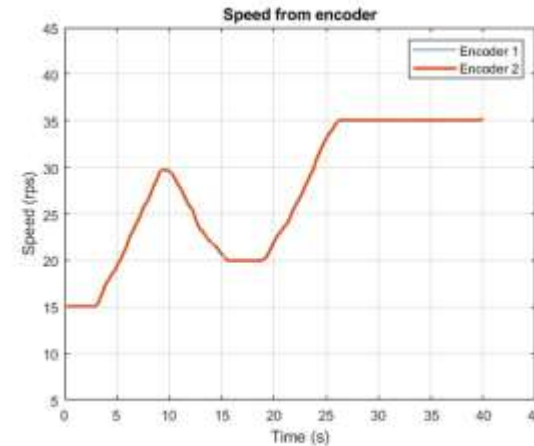
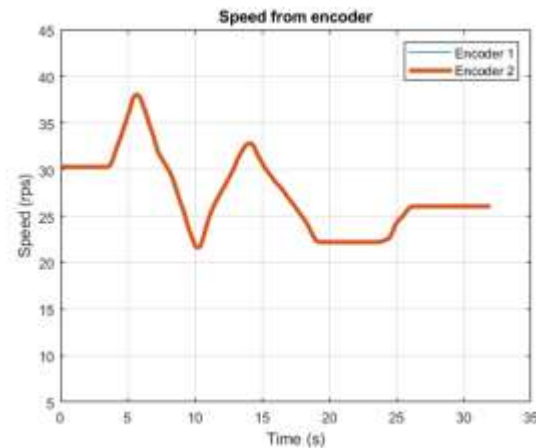
Healthy signals



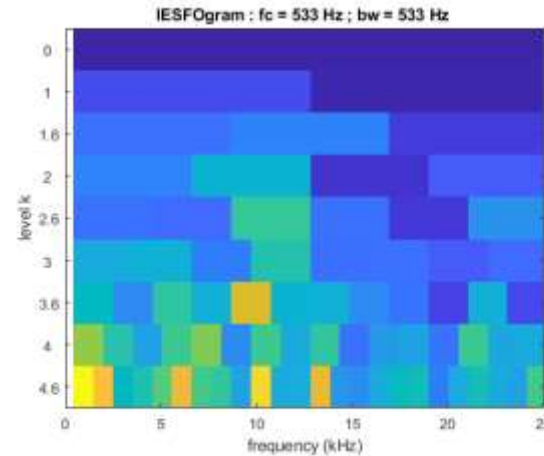
# Order Frequency Spectral Correlation



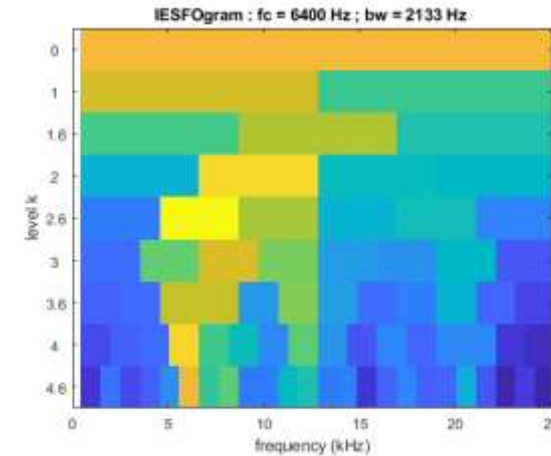
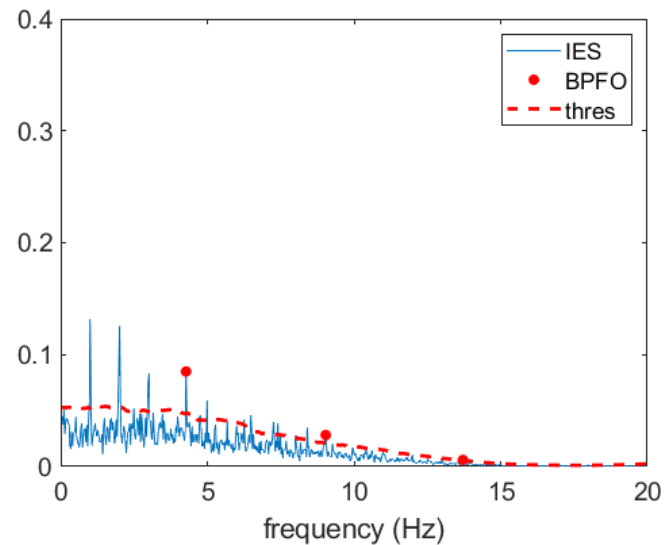
# Overall view on speed and OFCMS for healthy and damaged



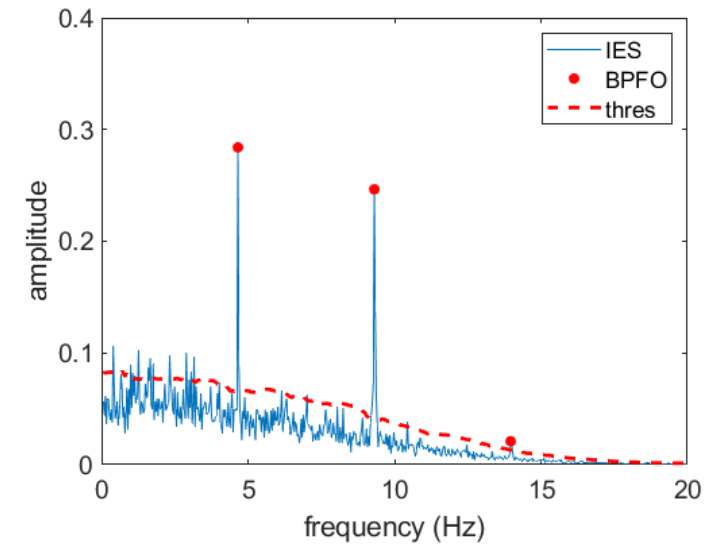
# IESFOgram targeted at BPFO



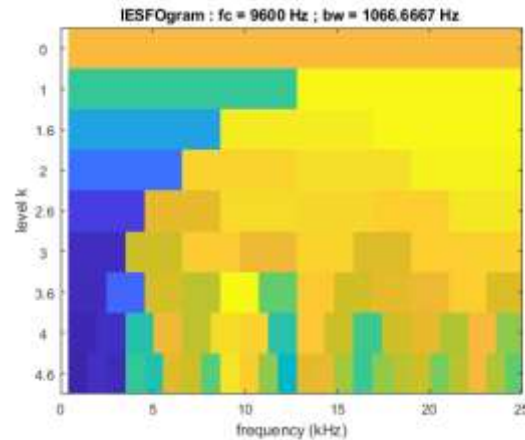
Healthy



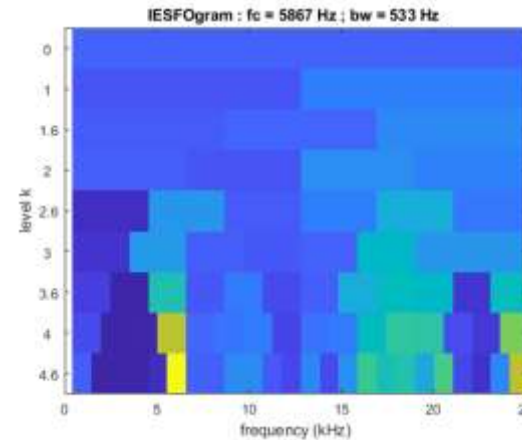
Damaged



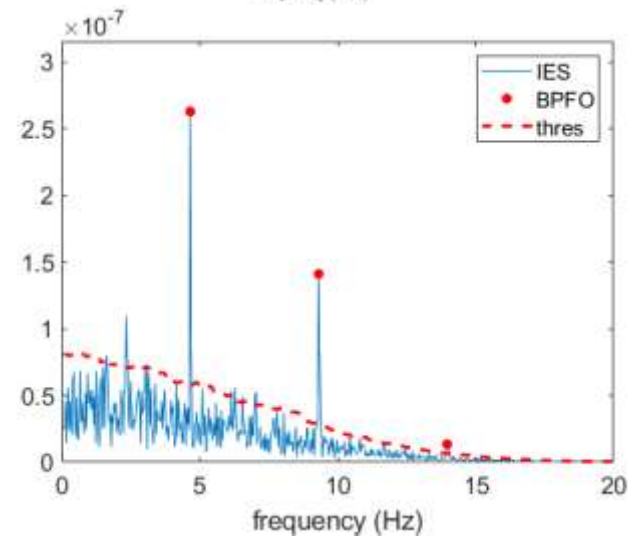
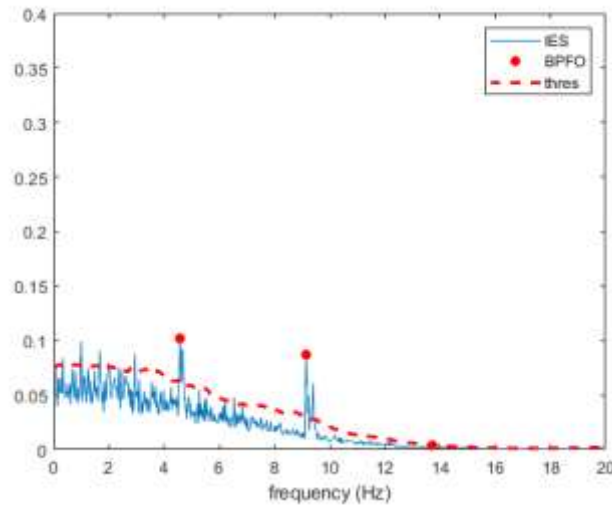
# Blind IESFOgram on damaged case



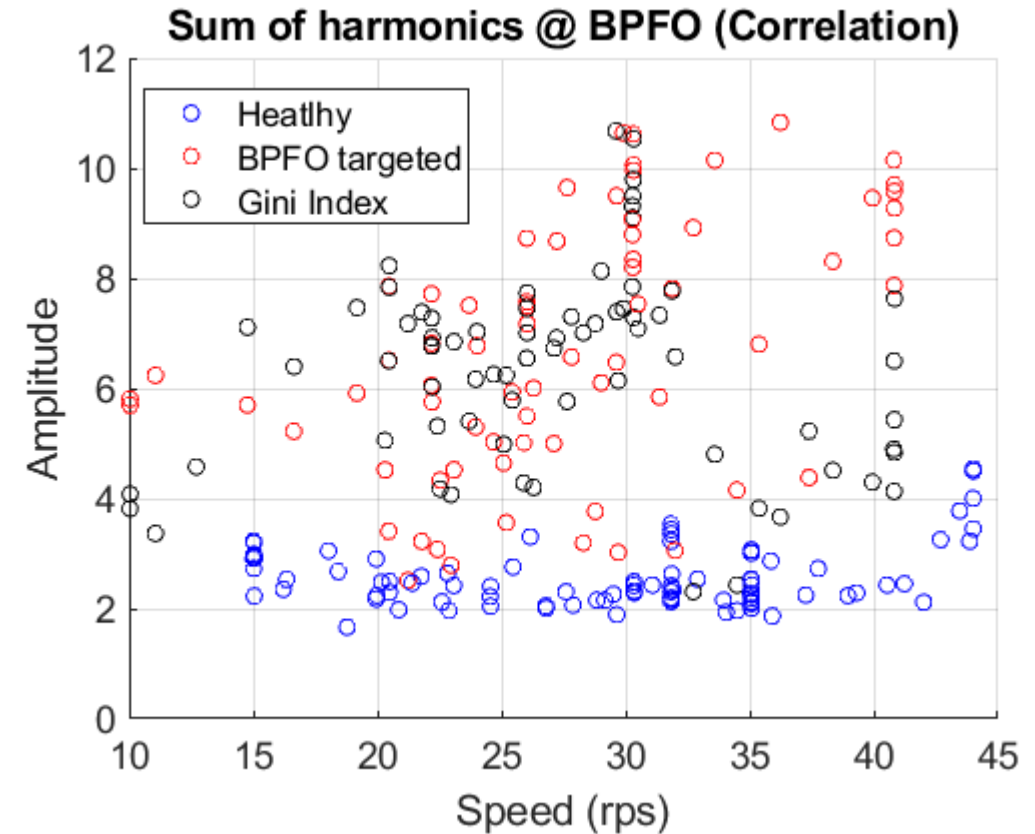
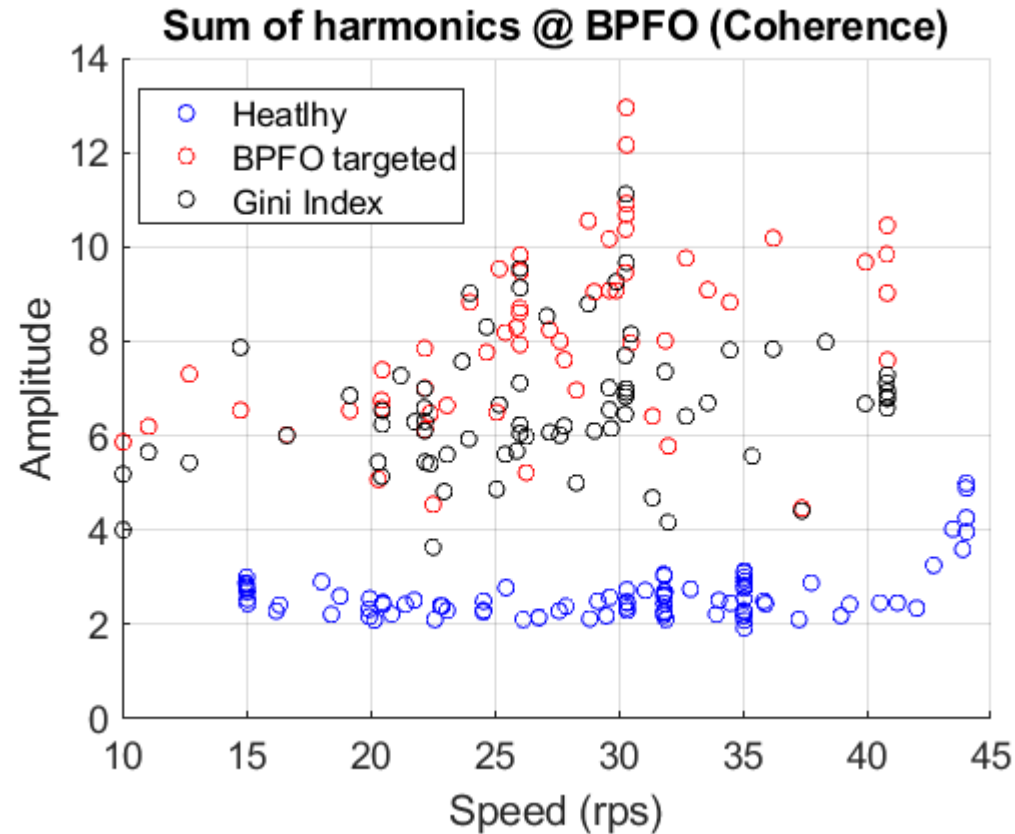
Kurtosis



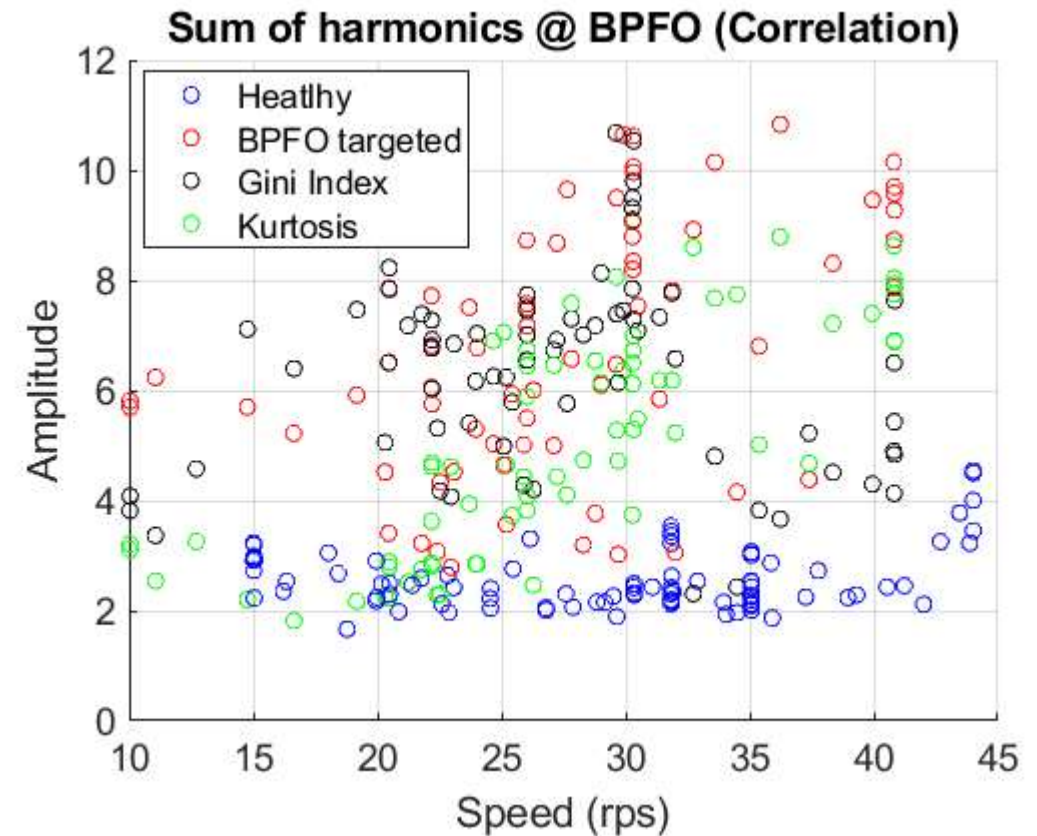
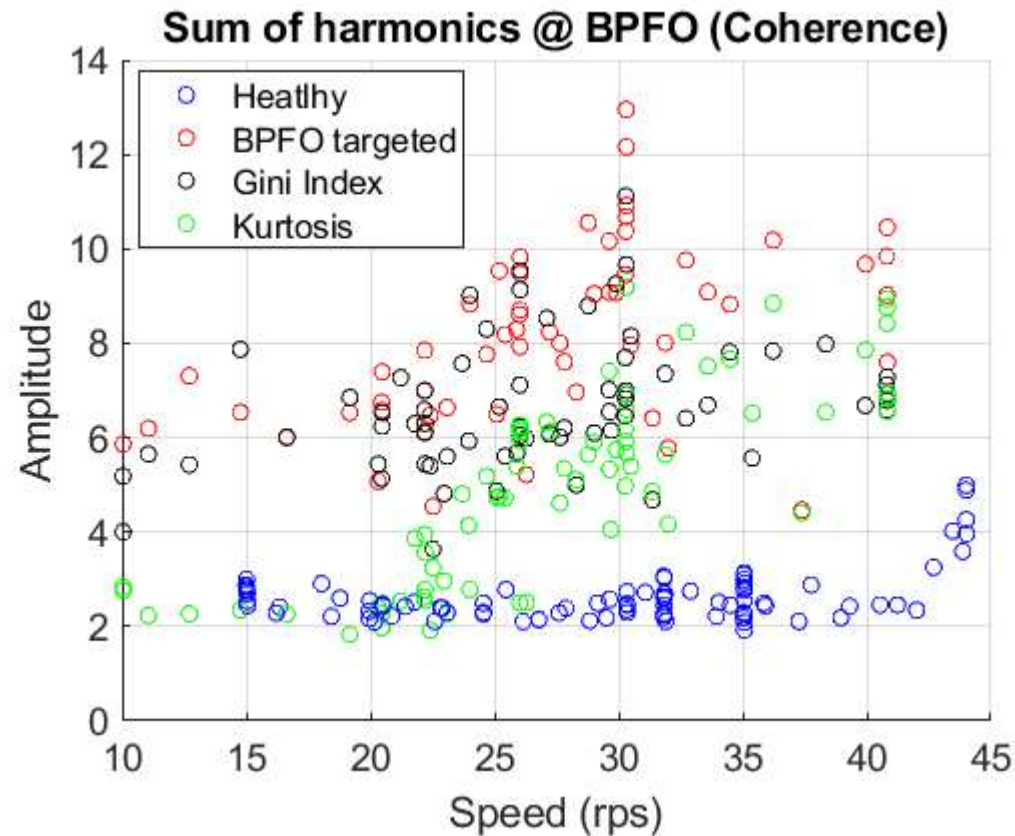
Gini Index



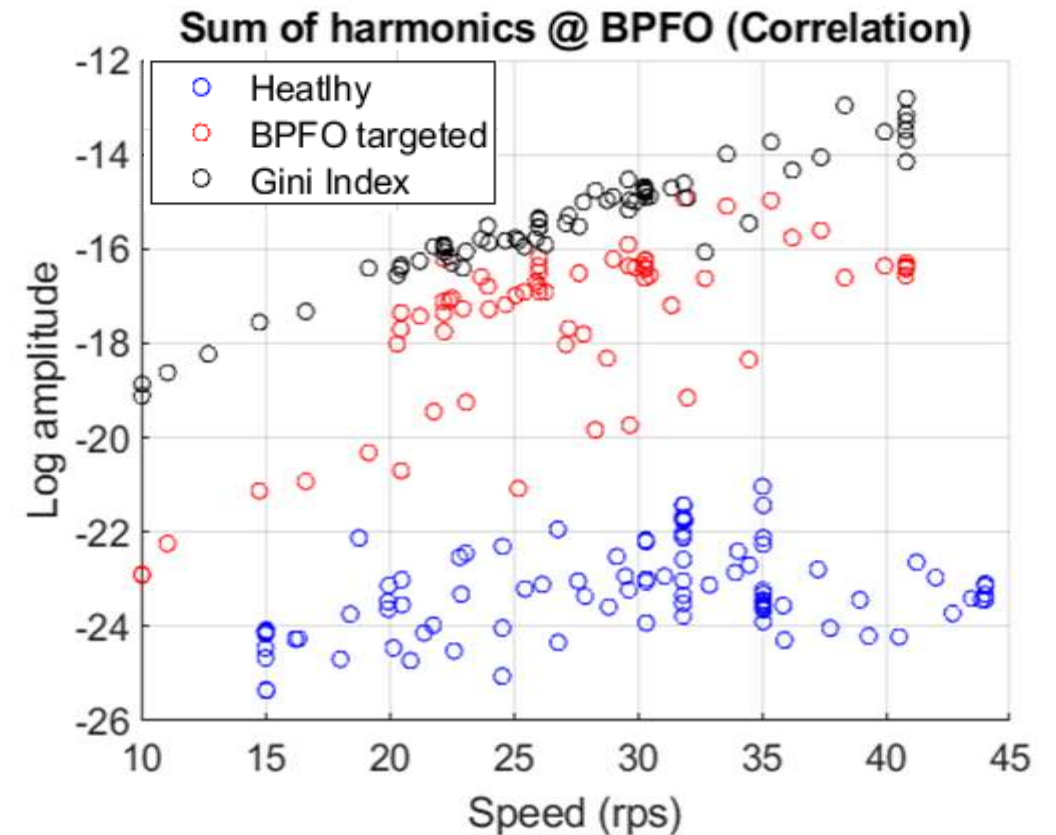
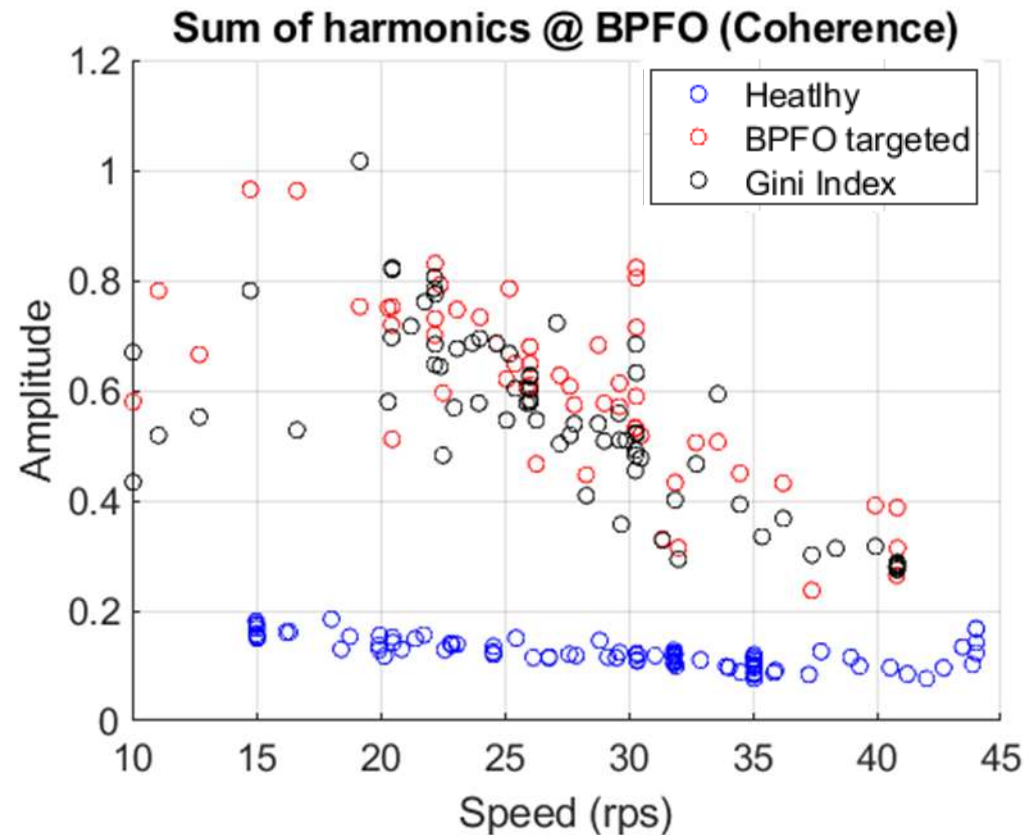
# Sum of harmonics at the BPFO (Normalized)



# Sum of harmonics at the BPFO (Normalized)



# Sum of harmonics at the BPFO (Absolute)



# Conclusions



# Conclusions

- A comparison of the performance of frequency-targeted features vs. blind feature for band selection has been performed.
- The features were extracted from the Cyclic Spectral Coherence in either Frequency-Frequency or Order-Frequency domain (IESFOgram)
- Vibration signals under steady and varying speed conditions were used to validate and research the features robustness.
- Frequency-Targeted feature provides the best performance, albeit kinematic knowledge of the rotating structure is necessary.
- Gini Index as a feature for band selection showed good performance for steady operating conditions and varying speed conditions at the DMMS-M drivetrain
- Other blind features did not show good diagnostic capabilities for the studied cases.

# Acknowledgement

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[alex.ricardomauricio@kuleuven.be](mailto:alex.ricardomauricio@kuleuven.be)



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# Thank you for your attention

Alexandre Mauricio

Department of Mechanical Engineering, Noise & Vibration Research Group, Belgium  
Dynamics of Mechanical and Mechatronics Systems, Flanders Make, Belgium  
[alex.ricardomauricio@kuleuven.be](mailto:alex.ricardomauricio@kuleuven.be)

