

Classification on Vibroarthrographic signals

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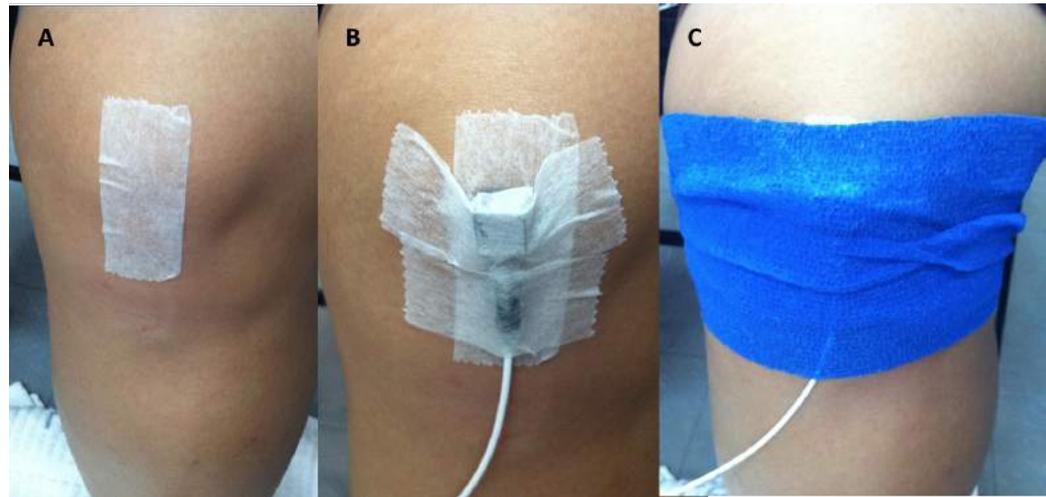
Chulalongkorn University

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Outline

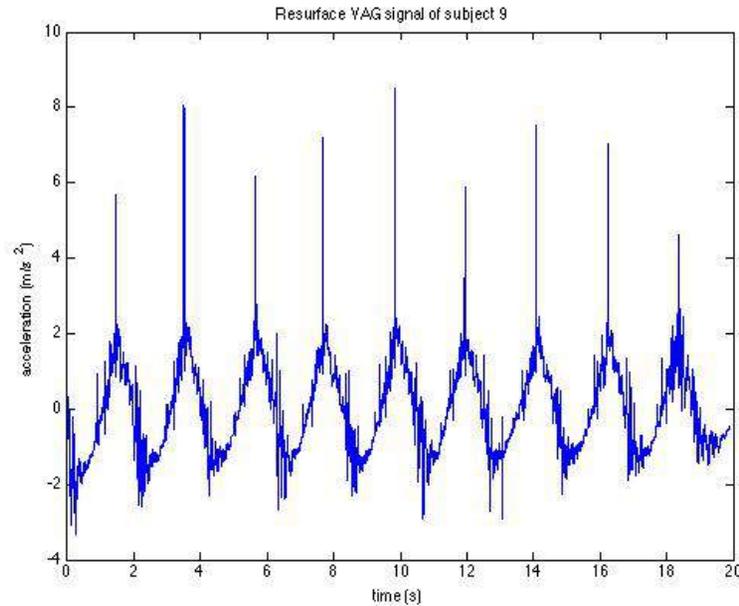
- Revisit VAG signals
- Noise removal with EEMD / DFA
- Overview on STFT results
- Challenge on classification problem

Recap on VAG Signal



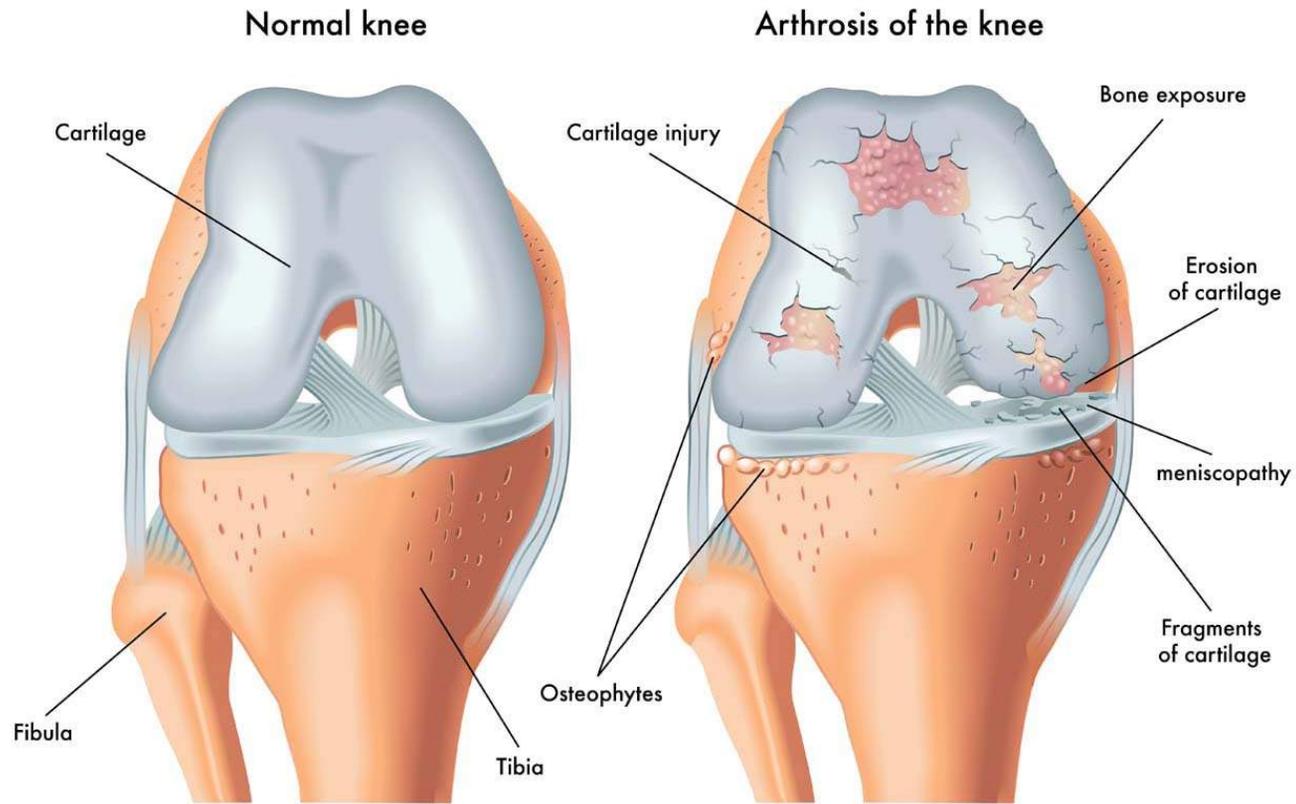
Demonstration of an attachment of an accelerometer.

An example of VAG signals



Three important characteristics:

1. Spike
2. Motion trend
3. Mid range frequency



Difference between abnormal and normal knees.

Photo by Renee Silvester

Classification on resurface and non-resurface

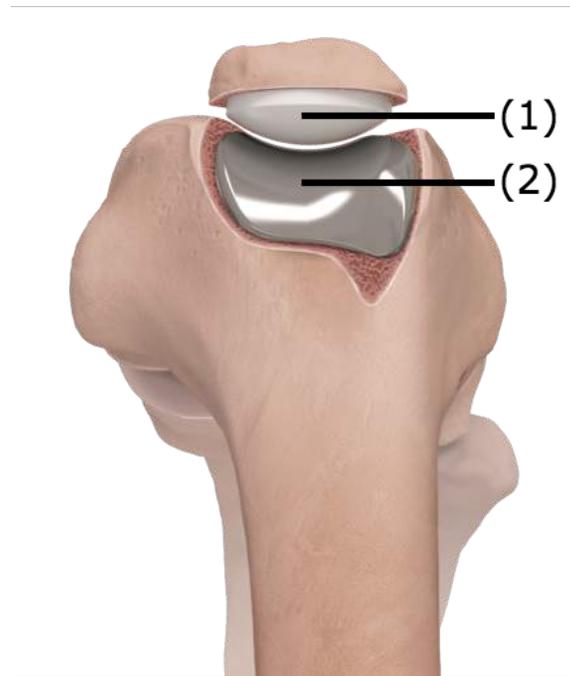
Goal classify normal and abnormal knee with VAG signal

Scope of work Since there are data from patients of two surgery methods available. We try to classify according to frequency and energy feature.

There are two class of data:

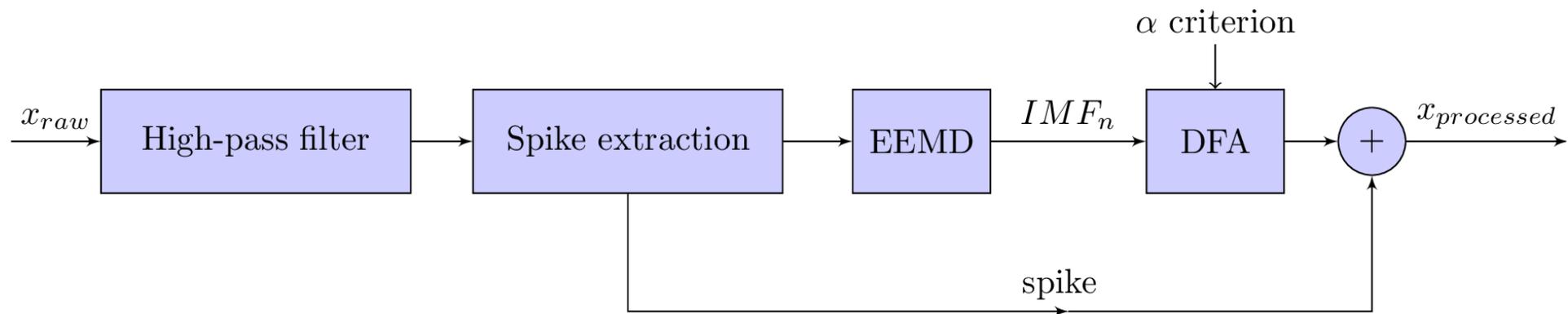
1. non-resurface: joint surface replacement (2)
2. resurface: joint surface (2) and patella replacement (1)

Assume: resurface \Rightarrow normal, non-resurface \Rightarrow abnormal



Label 1 shows an oval shape object used to replace a patella.

Processing overview



EEMD

(Ensemble Empirical Mode Decomposition)

It decomposes a signal into a set of signals called IMFs
(Intrinsic Mode Functions)

Why EEMD?

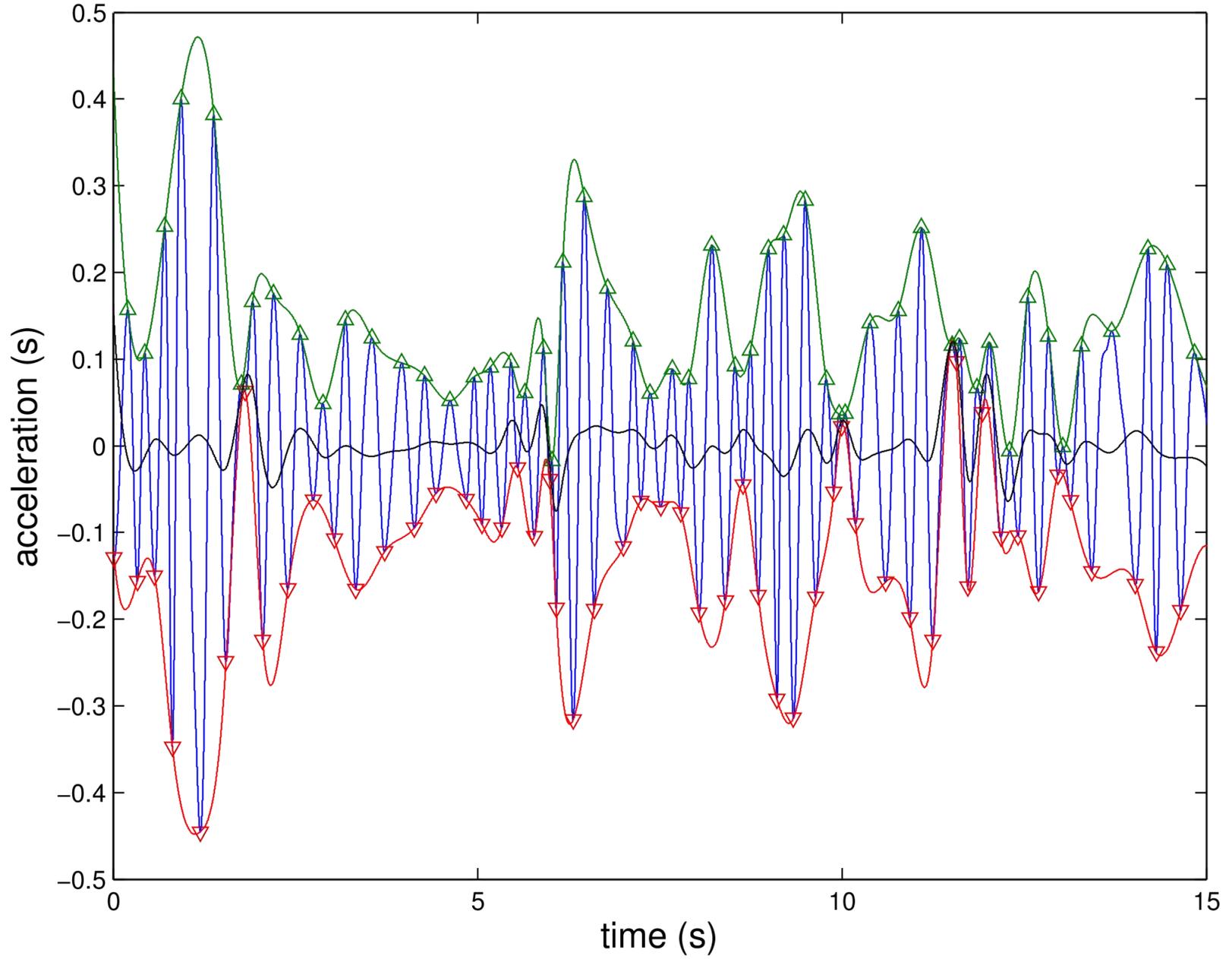
EEMD is common in physiological signals processing. Since VAG signals is non-stationary and noise are spread over the frequency band, EEMD is considered to be an appropriate choice.

How to obtain IMF

Sifting process: Interpolated upper and lower envelope and evaluate its mean. Then subtract this signal by this mean.

Repeat until reach (i) given iteration or (ii) signal has specific number of zero crossing.

Example spline of IMF#10



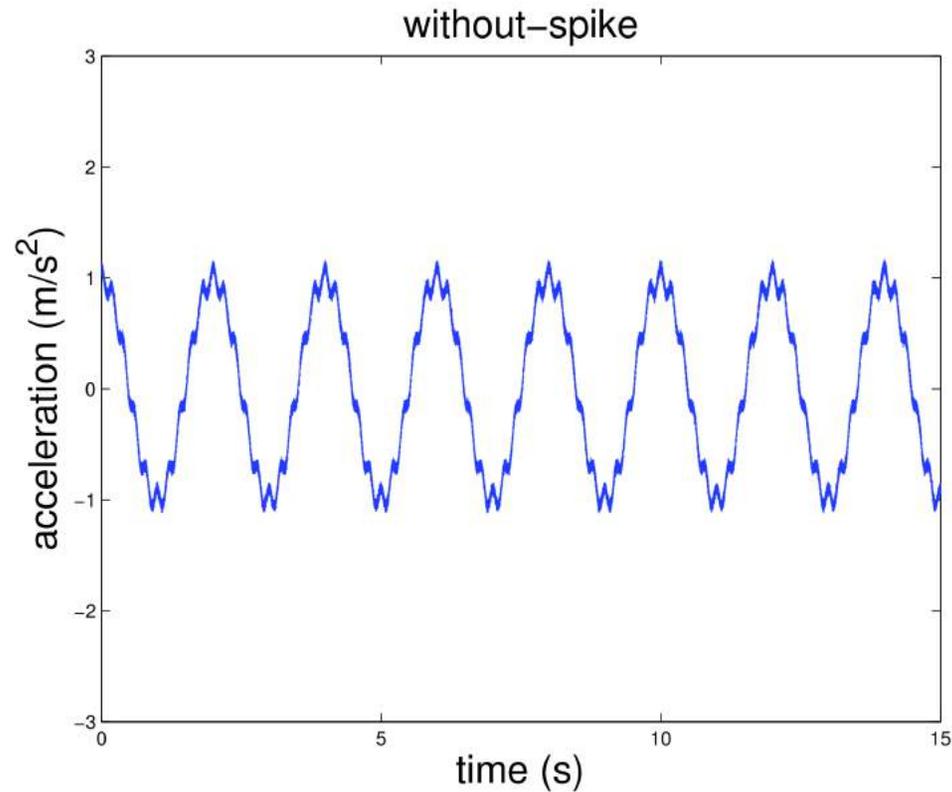
For iteration i of determining IMF c_i , we also obtain a residual r_i , the first iteration expression is given by

$$r_1 = x_{input} - c_1$$

Then the residual r_i will be the input of the next iteration.

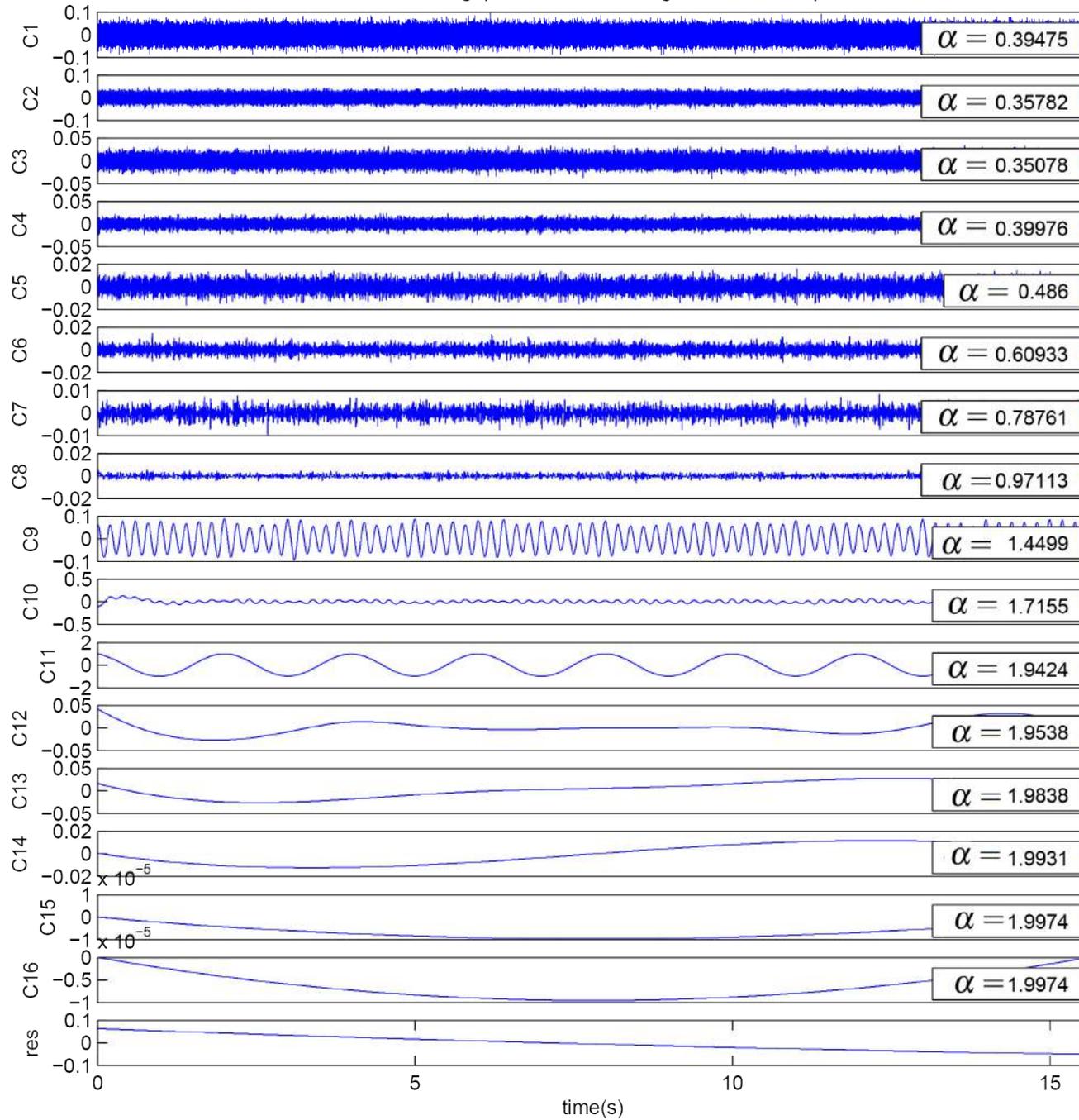
$$r_{i+1} = r_i - c_{i+1}$$

Example



$$y(t) = \cos(\omega t) + \cos(10\omega t) + n(t), n \sim \mathcal{N}(0, 0.1)$$

IMFs of the highpass filtered VAG signal of without-spike



DFA

(Detrended Fluctuation Analysis)

DFA is used to label whether an IMF is *correlated*.

$$C(s) = \mathbf{E}[x(t)x(t + s)] \approx \frac{1}{N - s} \sum_{t=1}^{N-s} x(t)x(t + s)$$

Correlation can be written in term of power law as

$$C(s) \propto s^{-\gamma}$$

Parameter γ can be indirectly estimated by Fluctuation function

$$F(s) = s^{1-\frac{\gamma}{2}} = s^{\alpha},$$

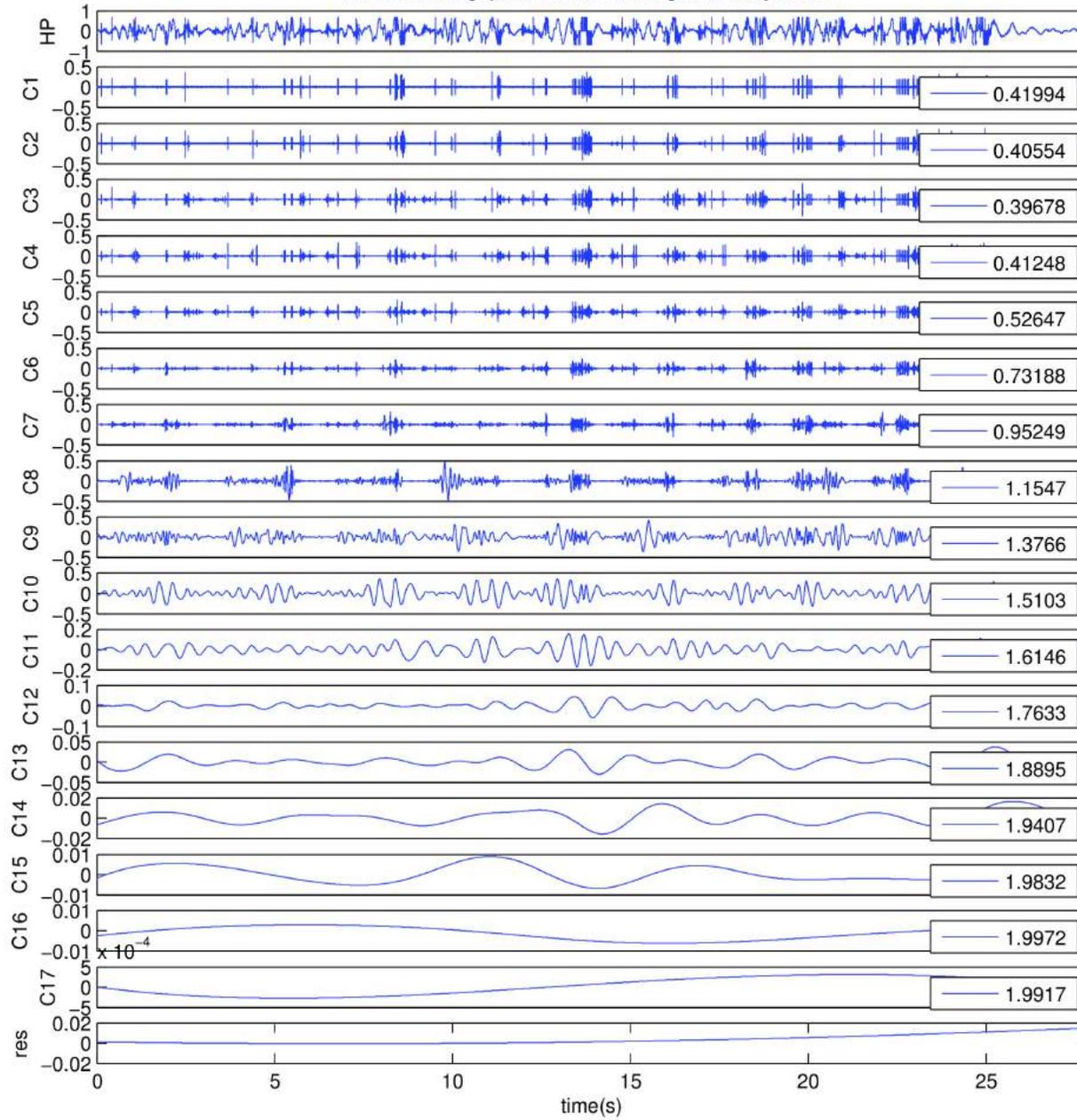
where s is a segmentation length of the signal.

Some important range of α are as followed:

- $\alpha < 0.5$ short-range correlated
- $\alpha = 0.5$ uncorrelated, white noise
- $\alpha > 0.5$ correlated

Normally, signal which $\alpha \leq 0.5$ is discard.

IMFs of the highpass filtered VAG signal of subject6-L1



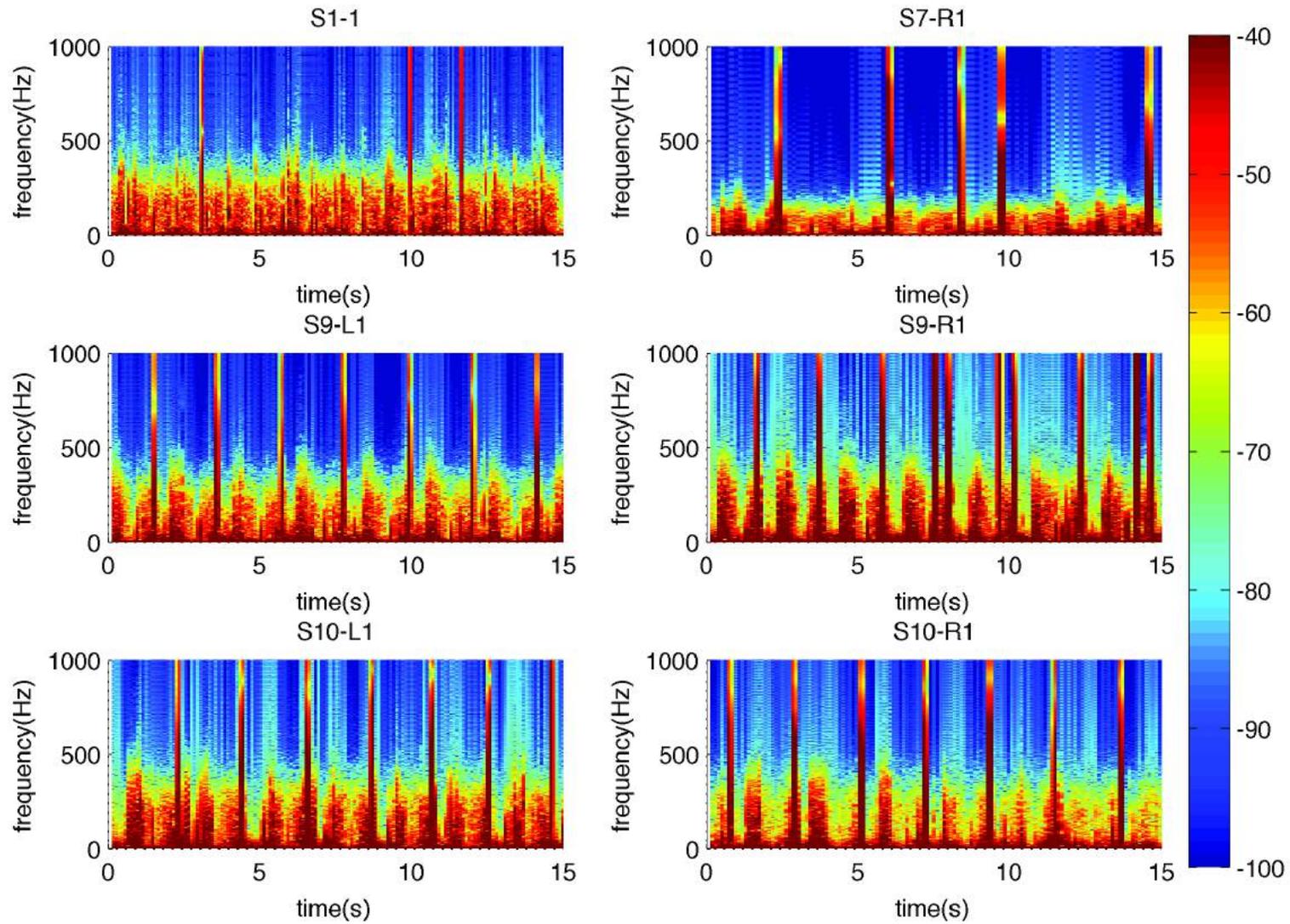
Short time Fourier transform (STFT)

STFT analyzes the signals into parts called window $w[n]$.
The STFT of a VAG signal $y[n]$ is given by

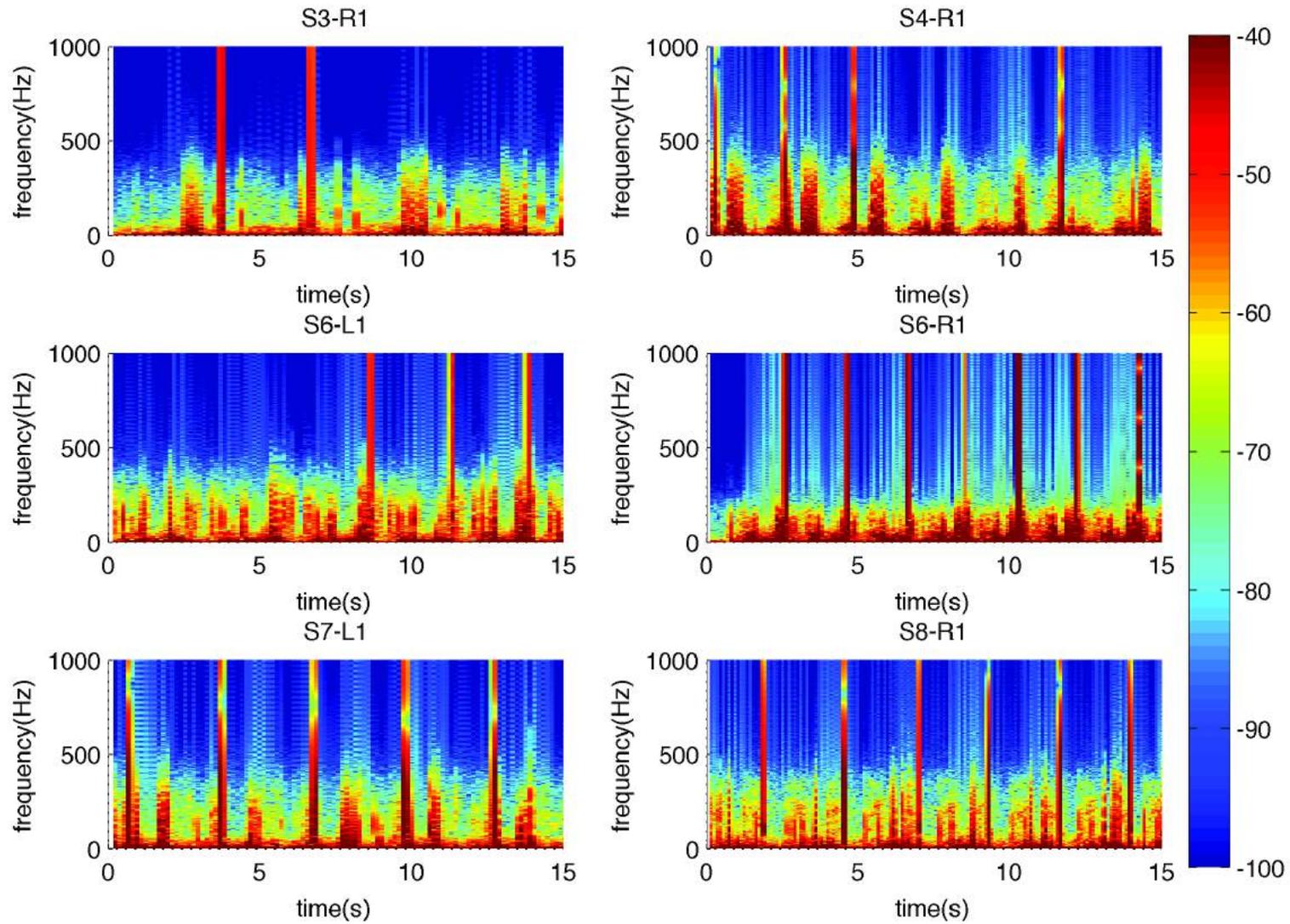
$$Y(m, \omega) = \sum_{n=m}^{L-m-1} y[n]w[n - m]e^{-j\omega n},$$

where $w[n]$ is a window function of length L

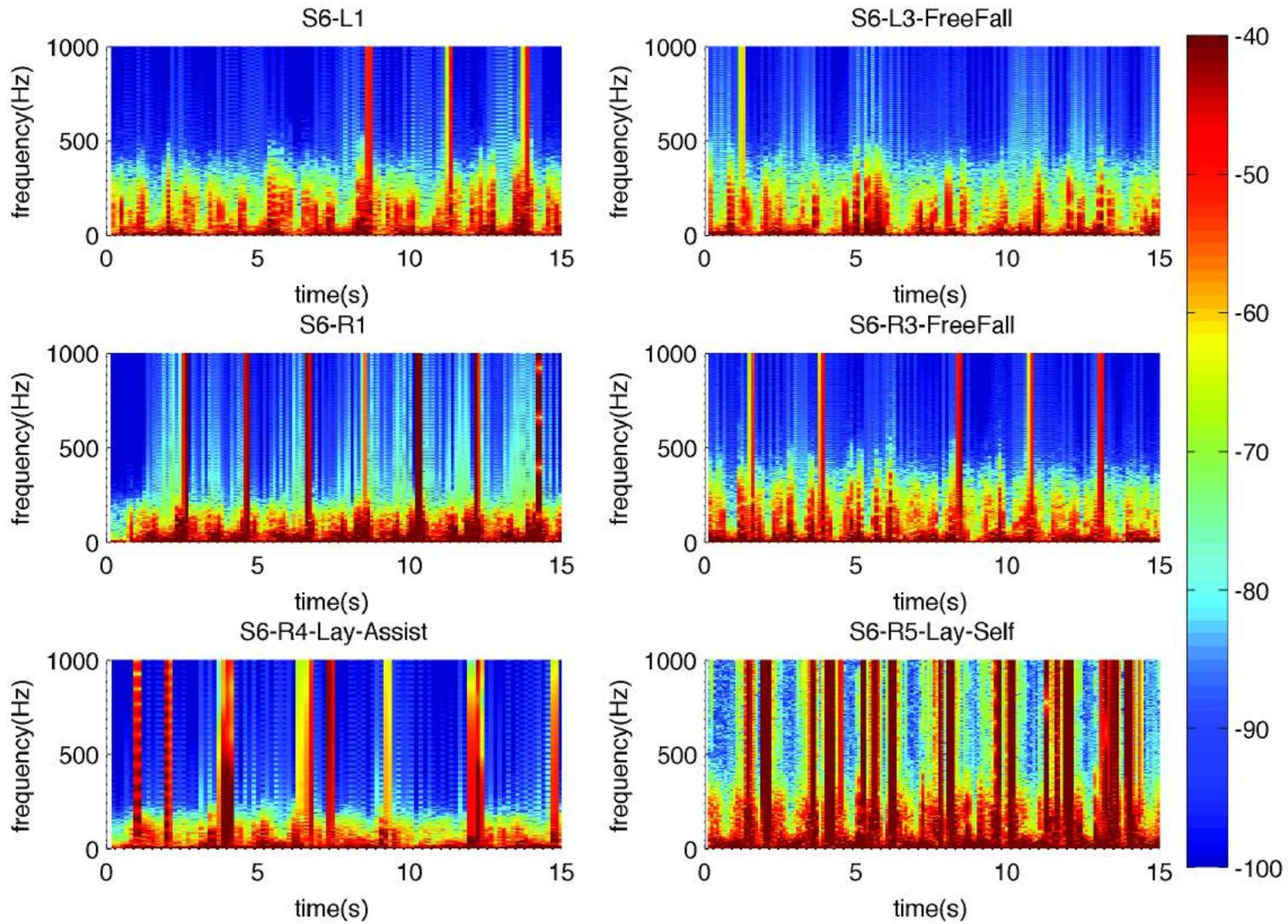
STFT results of resurface class



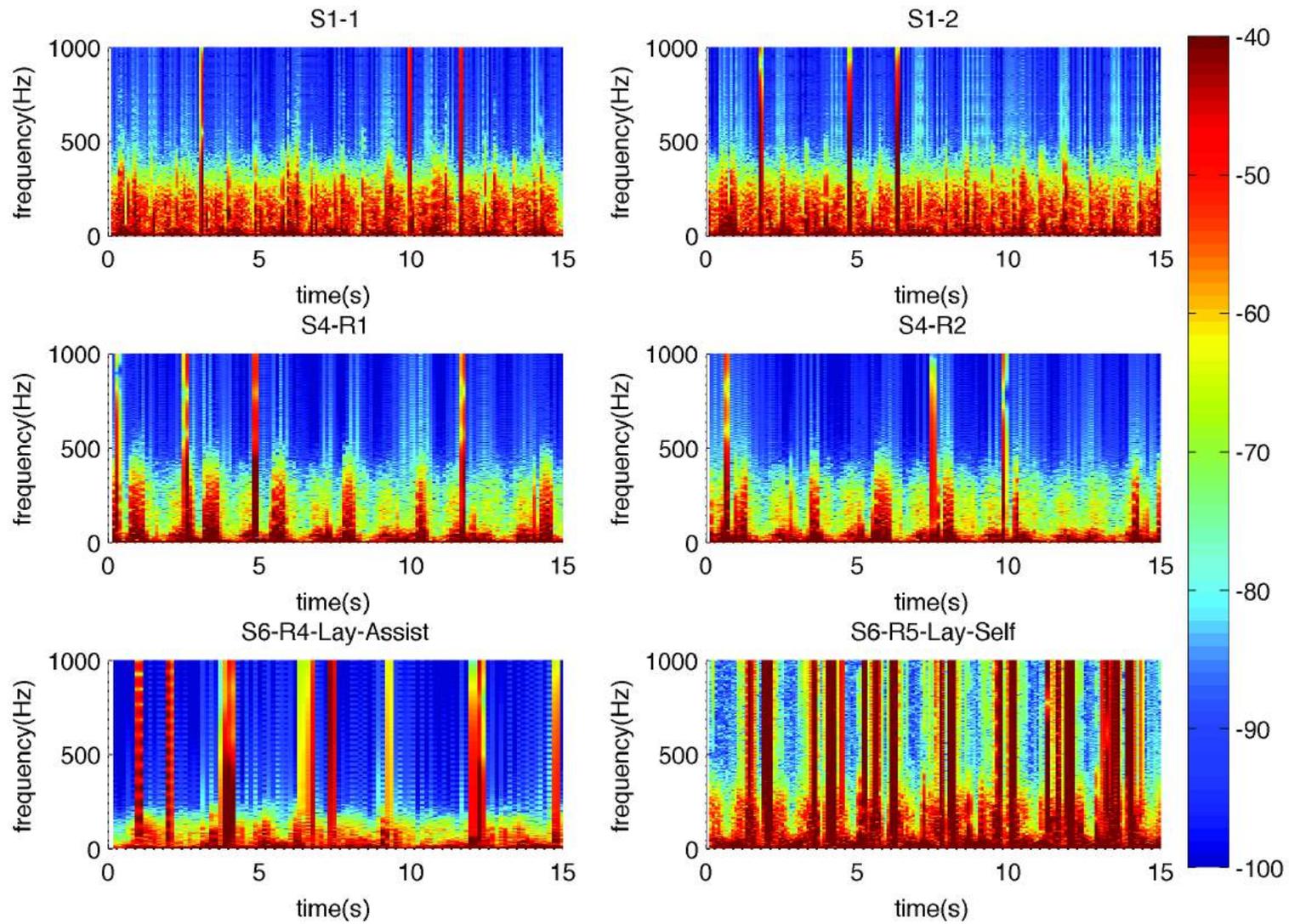
STFT results of non-resurface class



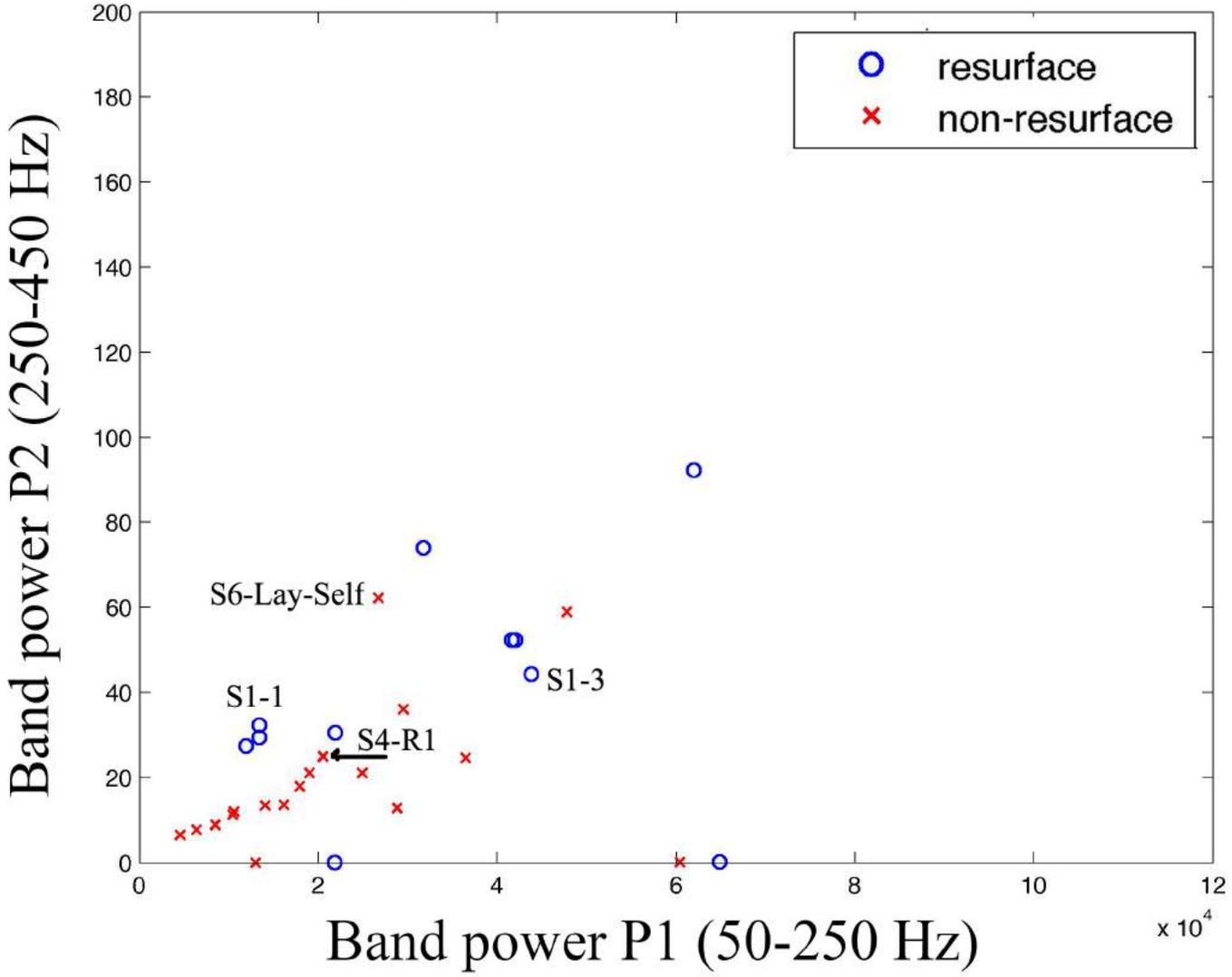
STFT results under different condition



STFT results of subject with sound



Signal power in different band



Conclusion

- Despite processing with EEMD / DFA, the resurface / non-resurface data set are not different.
- There are differences when measurement is carried out under different condition.
- Subjects with cracking sound has magnitude according to their severity.

Thank you
for your attention.

contact me via  : fangyibkk