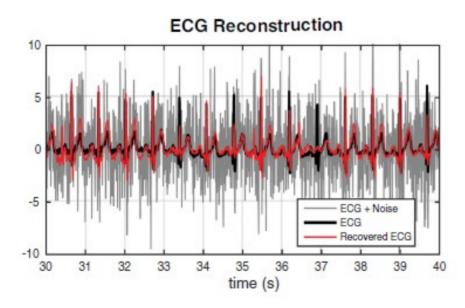
Introduction to Bispectral Analysis and High Order Correlations

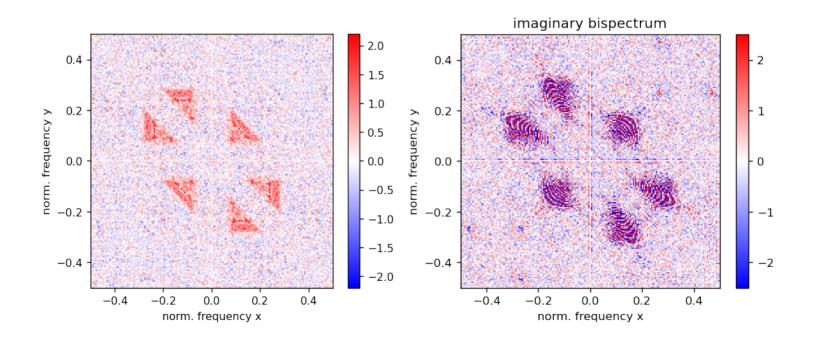
Jonas Meinel 01.04.2022

Motivation

 Reconstruction of broad-in-frequency signals with low SNR, e.g. electric/magnetic heart beat signals



How to read Bispectra? What do they mean?



Average

$$\mu = \langle x \rangle = \frac{1}{n} \sum_{i} x_{i}$$

Variance

$$var(x) = \sigma^2 = \langle (x - \mu)^2 \rangle$$

Skewness

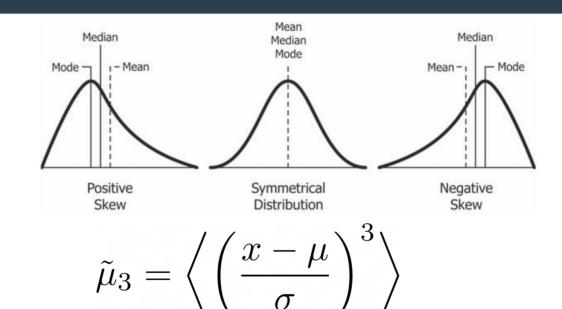
$$\tilde{\mu}_3 = \left\langle \left(\frac{x - \mu}{\sigma} \right)^3 \right\rangle$$

Kurtosis

Average

Variance

Skewness



Kurtosis

Average

$$\mu = \langle x \rangle = \frac{1}{n} \sum_{i} x_{i}$$

Variance

$$var(x) = \sigma^2 = \langle (x - \mu)^2 \rangle$$

Skewness

$$\tilde{\mu}_3 = \left\langle \left(\frac{x - \mu}{\sigma} \right)^3 \right\rangle$$

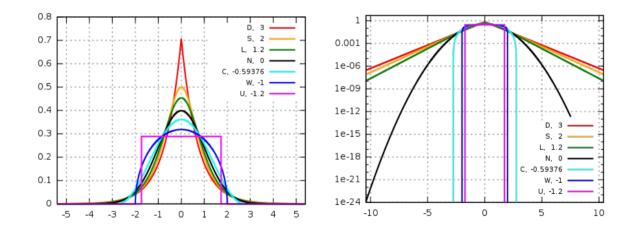
Kurtosis

$$\tilde{\mu}_4 = \left\langle \left(\frac{x - \mu}{\sigma} \right)^4 \right\rangle = \left\langle \operatorname{var} \left(\left(\frac{x - \mu}{\sigma} \right)^2 \right) \right\rangle + 1$$

Average

Variance

Skewness



$$\tilde{\mu}_4 = \left\langle \left(\frac{x - \mu}{\sigma} \right)^4 \right\rangle = \left\langle \operatorname{var} \left(\left(\frac{x - \mu}{\sigma} \right)^2 \right) \right\rangle + 1$$

Higher Order Correlations (HOC)

- 1st order
- 2nd order
- 3rd order

$$C_2(\tau) = \langle x(t)x(t+\tau)\rangle$$

 $C_1 = \langle x(t) \rangle$

• 4th order

$$C_3(\tau_1, \tau_2) = \langle x(t)x(t + \tau_1)x(x + \tau_2) \rangle$$

$$C_4(\tau_1, \tau_2, \tau_3) = \langle x(t)x(t + \tau_1)x(t + \tau_2)x(t + \tau_3) \rangle$$

HOC and Higher Order Spectra (HOS)

Power spectrum

$$S_{xx}(f) = \int_{-\infty}^{\infty} C_2(\tau) e^{-i2\pi f \tau} d\tau = \mathcal{F}_{C_2}(f),$$

Bispectrum

$$B_x(f_1, f_2) = \int \int d\tau_1 d\tau_2 C_3(\tau_1, \tau_2) e^{-i2\pi(f_1\tau_1 + f_2\tau_2)}.$$

- Trispectrum: ...
 - → Why Higher order Spectra are interesting?

Motivation of Bispectra

Most important identity of Bispectra

$$B_x(f_1, f_2) = \mathcal{F}_x(f_1) \mathcal{F}_x(f_2) \mathcal{F}_x^{-1}(f_1 + f_2),$$

- Sensitive to non-linear processes $y = ax^2 + bx$
- Conserving the relative phase of signals $\;B_x(f_1,f_2)\in\mathbb{C}^2$
- "Gaussian noise free" $C_{n>2}=0$

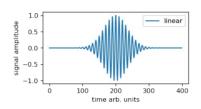
Motivation of Bispectra

Very important identity of Bispectrum

$$B_x(f_1, f_2) = \mathcal{F}_x(f_1) \mathcal{F}_x(f_2) \mathcal{F}_x^{-1}(f_1 + f_2),$$

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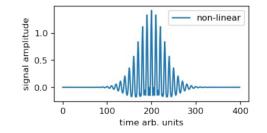
Original signal

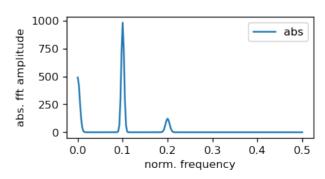


Signal after non linear process

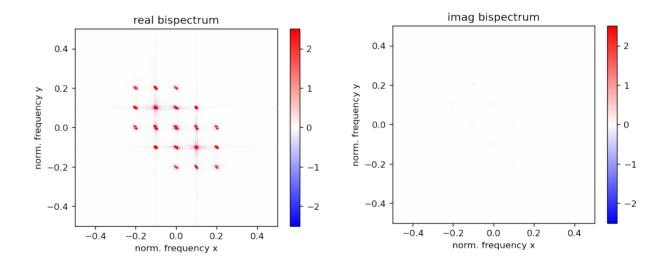
$$y = ax^2 + bx$$

 Power spectrum of the signal after non-linear process

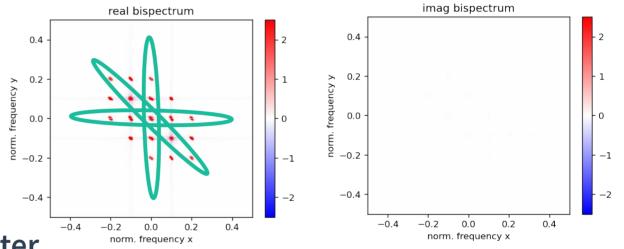




- Quadratic Phase coupled (QPC) component
 - → real peaks

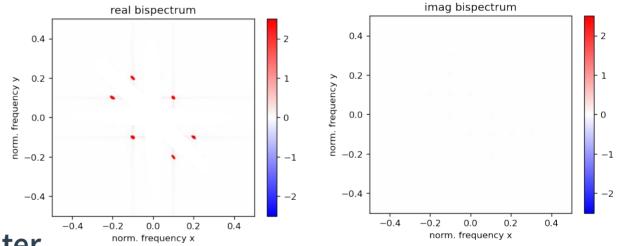


- Quadratic Phase coupled (QPC) component
 - → real peaks
- The peaks involving f(0)
 -0.4
 are trivial → high pass filter



$$B_x(f_1, f_2) = \mathcal{F}_x(f_1) \mathcal{F}_x(f_2) \mathcal{F}_x^{-1}(f_1 + f_2),$$

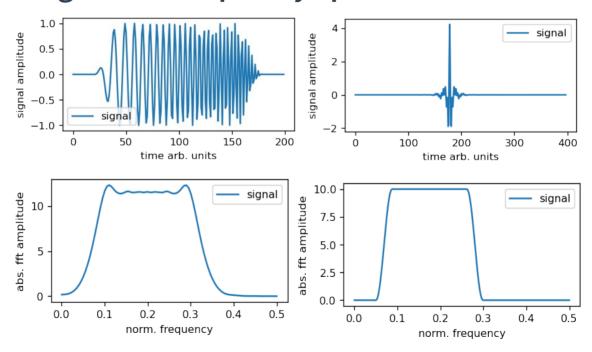
- Quadratic Phase coupled (QPC) component
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$$B_x(f_1, f_2) = \mathcal{F}_x(f_1) \mathcal{F}_x(f_2) \mathcal{F}_x^{-1}(f_1 + f_2),$$

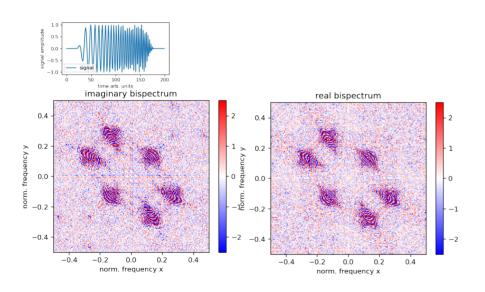
Example 2: Square Signals

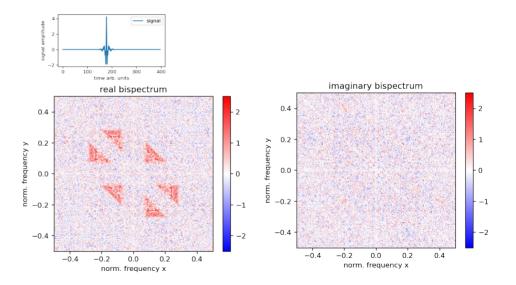
• Two rectangular signals in frequency space



Example 2: Square Signals

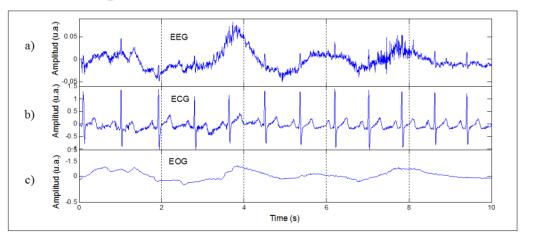
• The bispectrum is sensitive to the complex phase

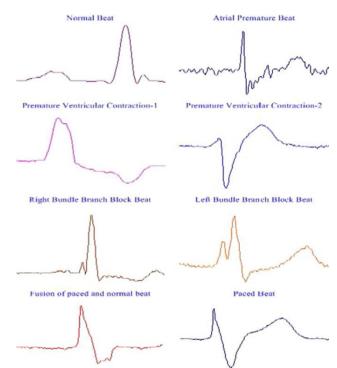




Human Electrical Signals

- Common cardiac arrhythmias (ECG)
- EEG signals

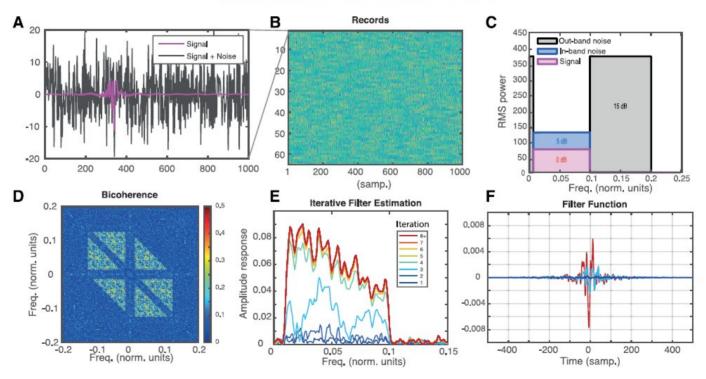




Lanata, Antonio, et al. Expert Systems with Applications 38.6 (2011): 6798-6804. Correa, M. Agustina Garcés, and Eric Laciar Leber. Adaptive filtering applications 34 (2011): 1-26.

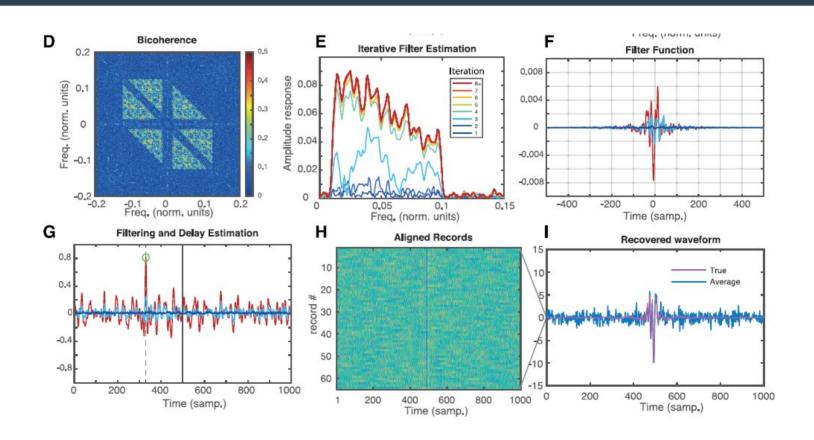
Example 3: Reconstructing a Filter for Arbitrary Signals



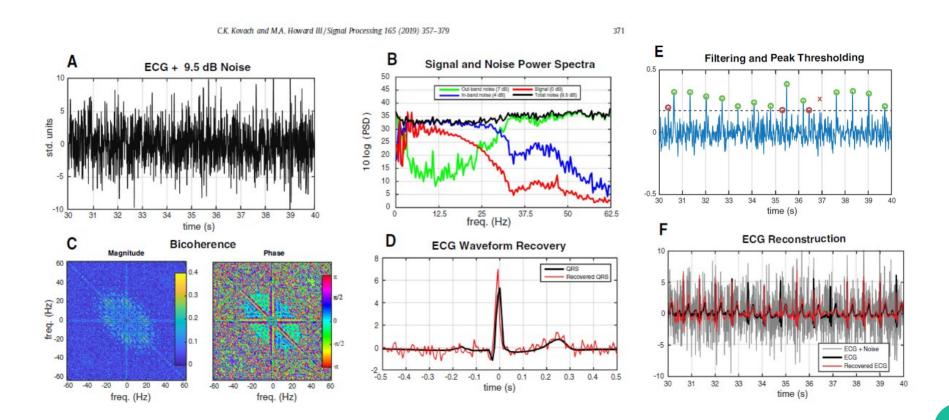


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Example 3: Reconstructing a Filter for Arbitrary Signals



Example 4: Reconstructing a Filter for Bio Signals



Summary of Bispectral analysis

Very important identity of Bispectrum

$$B_x(f_1, f_2) = \mathcal{F}_x(f_1) \mathcal{F}_x(f_2) \mathcal{F}_x^{-1}(f_1 + f_2),$$

- Sensitive to non-linear processes $y = ax^2 + bx$
- Conserving the relative phase of signals $\;B_x(f_1,f_2)\,\in\mathbb{C}^2$
- "Gaussian noise free" $C_{n>2}=0$

Acknowledgements

Jörg Wrachtrup & Vadim Vorobyov & the team at PI3



