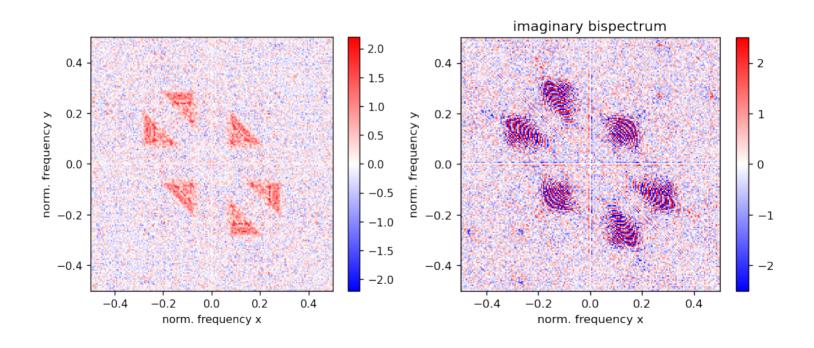
# Introduction to Bispectral Analysis and High Order Correlations

Jonas Meinel 10.12.2021

## 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

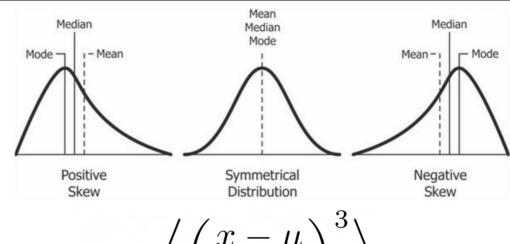
$$\tilde{\mu}_3 = \left\langle \left( \frac{x - \mu}{\sigma} \right)^3 \right\rangle$$

Kurtosis

Average

Variance

Skewness



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Kurtosis

Average

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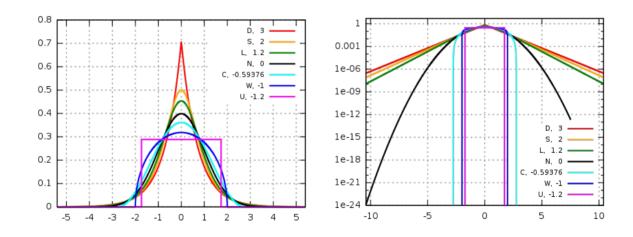
Kurtosis

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

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)**

- 1<sup>st</sup> order
- 2<sup>nd</sup> order
- 3<sup>rd</sup> order

• 4<sup>th</sup> order

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

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

ler 
$$C_3( au_1, au_2) = \langle x(t)x(t+ au_1)x(x+ au_2) 
angle$$
  $C_4( au_1, au_2, au_3) = \langle x(t)x(t+ au_1)x(t+ au_2)x(t+ au_3) 
angle$ 

## **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**

Very 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**

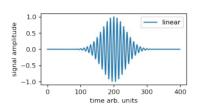
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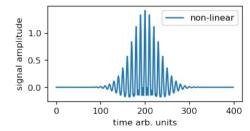
- Sensitive to non-linear processes  $y=ax^2+bx$  Conserving the relative phase of signals  $B_x(f_1,f_2)\in\mathbb{C}^2$ 
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## **Example 1: Non Linear Process**

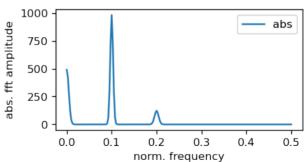
Original signal



Signal after non linear process



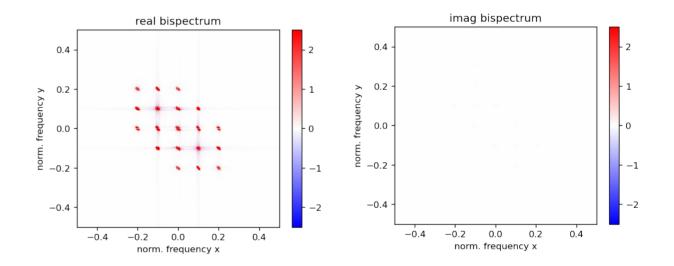
Power spectrum of the signal



## **Example 1: Non Linear Process**

- Quadratic Phase coupled (QPC) component
  - → real peaks
- Let's compare to

   a signal not from a
   non linear process
   with the same PSD

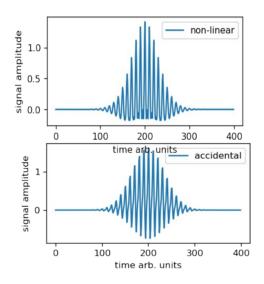


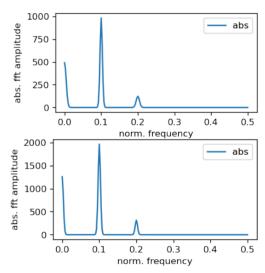
### QPC vs. accidental matched

Signal with same PSD,
 but 2f is phase shifted
 → accidentally matched

Top: QPC

Bottom: Accident

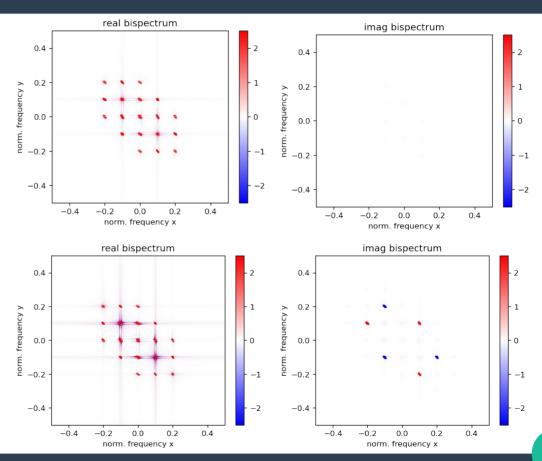




## QPC vs. accidental matched - bispectrum

Top: QPC

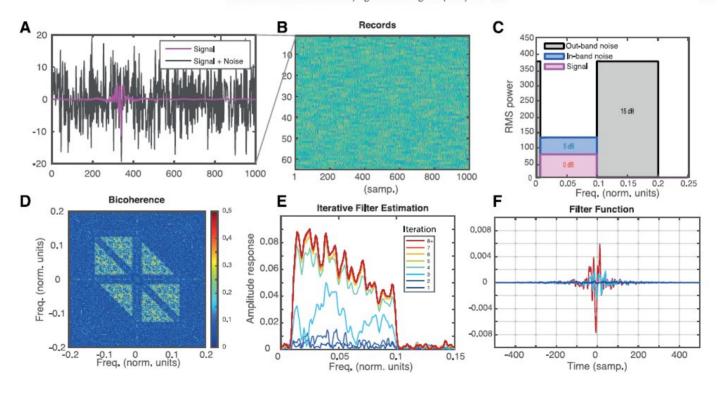
Bottom: Accident



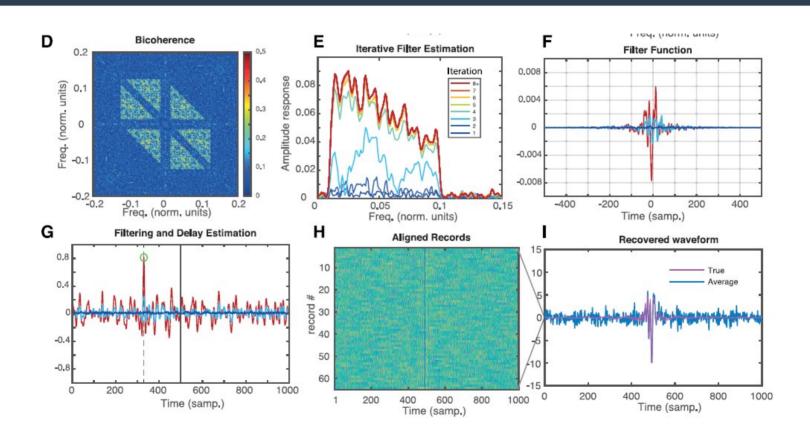
## **Example 2: Reconstructing a Filter for Arbitrary Signals**

C.K. Kovach and M.A. Howard III / Signal Processing 165 (2019) 357-379

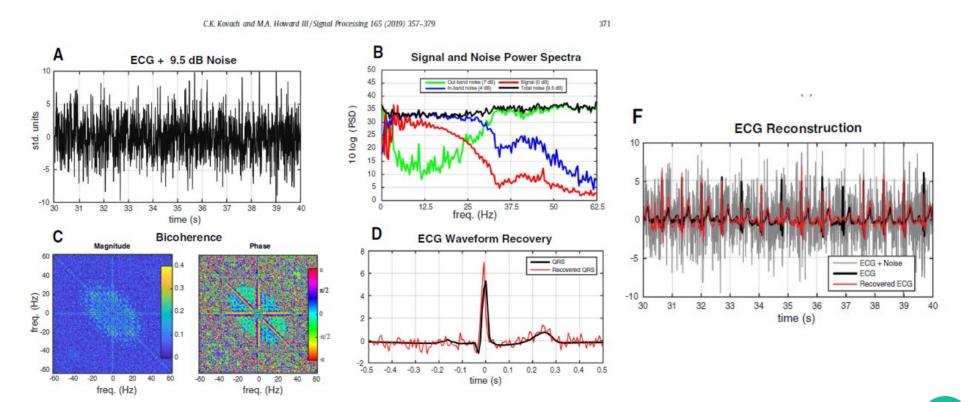




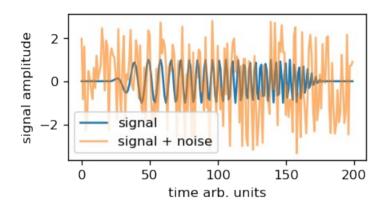
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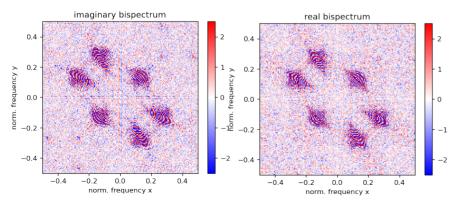


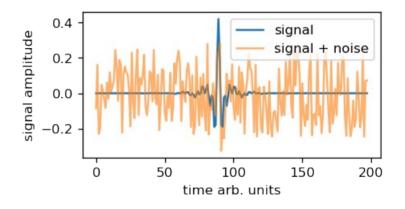
## **Example 2: Reconstructing a Filter for Bio Signals**

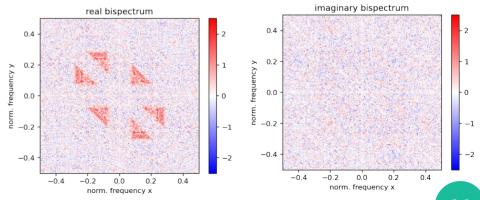


## **Other Examples:**









### **Summary of Bispectral analysis**

Very important identity of Bispectrum

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