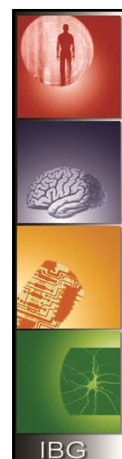


Signal & Data Analysis in Neuroscience
2016
Spectral Analysis

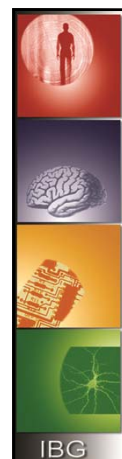
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Outline – Frequency domain

- ☑ Introduction
- ☑ Fourier Transform
- ☑ Sampling Theory
- ☑ Systems
- ☑ Filters
- **Spectral Analysis**

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Correlation & Convolution





- Correlation

$$R_{g,h}(t) = \text{Corr}(g, h)_t = \sum_{i=-\infty}^{\infty} g_i h_{t+i}$$

$$R_{g,h}(t) = \text{Corr}(g, h)_t = \text{Corr}(h, g)_{-t} = R_{h,g}(-t)$$
- Convolution

$$(g * h)_t = \sum_{i=-\infty}^{\infty} g_{t-i} h_i$$

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Parseval's theorem

The Fourier transform is unitary \rightarrow the sum (or integral) of the square of a function is equal to the sum (or integral) of the square of its transform.

Continuous Fourier transform $\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$

Discrete Fourier transform $\sum_{n=0}^{N-1} |x[n]|^2 = \sum_{k=0}^{N-1} |X[k]|^2$

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Signal energy





- The energy spectral density describes how the energy (or variance) of a signal or a time series is distributed with frequency.

$$\Phi(\omega) = \left| \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt \right|^2 = \frac{F(\omega) F^*(\omega)}{2\pi}$$

$$\Phi(\omega) = \left| \frac{1}{\sqrt{2\pi}} \sum_{n=-\infty}^{\infty} f_n e^{-i\omega n} \right|^2 = \frac{F(\omega) F^*(\omega)}{2\pi}$$

(only for finite energy signals)

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Signal Power

$$Power = \frac{Energy}{Time}$$





$$S = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T |x(t)|^2 dt$$

$$S = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-\infty}^{\infty} |X_T(f)|^2 df$$

Equal By Parseval's Theorem

$$\hat{S}[f] = |DFT\{x[n]\}|^2 = \left| \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{\left(\frac{-j2\pi n f}{N}\right)} \right|^2$$





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Wiener-Khinchin theorem

- The power spectrum is the Fourier transform of the auto-correlation function
- Power spectrum $S(f) = \int_{-\infty}^{\infty} R(\tau) e^{-j2\pi f\tau} d\tau$
- Autocorrelation $R(\tau) = \int_{-\infty}^{\infty} S(f) e^{j2\pi f\tau} df$

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Power spectral density





- Amount of power per unit (density) of frequency (spectral) as a function of the frequency

Power Spectrum	Spectral Density
$S_{x,x}(\omega) = \sum_{m=-\infty}^{\infty} R_{x,x}(m) \cdot e^{-jom}$	$P_{x,x}(\omega) = \frac{S_{x,x}(\omega)}{2\pi}$
$S_{x,x}(f) = \sum_{m=-\infty}^{\infty} R_{x,x}(m) \cdot e^{\frac{-2\pi fm}{f_s}}$	$P_{x,x}(f) = \frac{S_{x,x}(f)}{f_s}$

Average power

$$\bar{P}(\omega_1, \omega_2) = \int_{\omega_1}^{\omega_2} P_{x,x}(\omega) d\omega = \int_{-\omega_2}^{-\omega_1} P_{x,x}(\omega) d\omega = 2 \cdot \int_{\omega_1}^{\omega_2} P_{x,x}(\omega) d\omega$$





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Spectrum Estimation - Problems

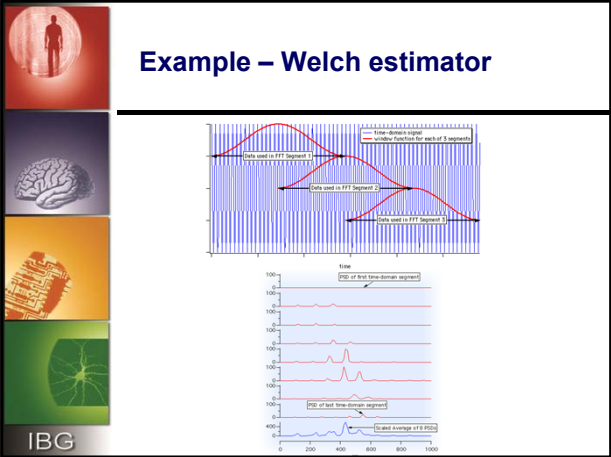
- Leakage problems and side lobes.
- Increased length of signal leads to increase in number of discrete frequency but not to increased accuracy at each frequency.





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Spectrum Estimation Methods





- Non parametric estimators
- Correlogram estimators
- Parametric estimators
- Subspace estimators



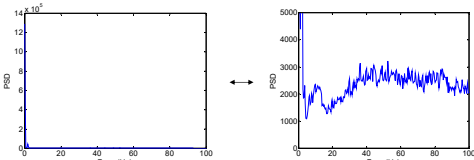
Things we find in the spectrum:

- DC
- White noise
- Harmonics










Direct Current (DC)

- DC is transformed into energy at frequency 0.
- Unless DC is removed the power it adds is typically huge compared to all other frequencies.

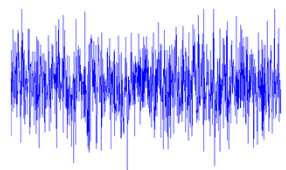


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White noise

- White noise is a random signal (or process) with a flat power spectral density.

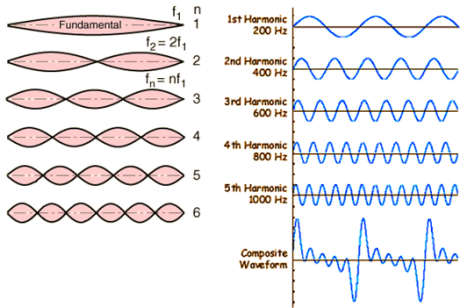


- Zero correlation at $t \neq 0$.
- There are also pink & red/brown noises.

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Harmonics



n	Frequency (Hz)	Waveform Description
1	200 Hz	1st Harmonic
2	400 Hz	2nd Harmonic
3	600 Hz	3rd Harmonic
4	800 Hz	4th Harmonic
5	1000 Hz	5th Harmonic
6	1200 Hz	6th Harmonic

Composite Waveform





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Do we have harmonies?

- Many biological processes lead to the formation of harmonies.
- Any square wave (for example a sudden change in a parameter) is transformed to multiple harmonies.
- The multiple frequencies may describe the same underlying process.

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Temporal & spectral resolution

- Using **windowed estimation** (Welch/Bartlett) leads to a temporal / spectral resolution tradeoff.
- For a recording of **T** seconds sampled at **s** samples/sec and assessed using a **w** sample window:

Number of windows: $\frac{T \cdot s}{w}$

Spectral resolution (Δf): $\frac{s}{w}$





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Relative & absolute power

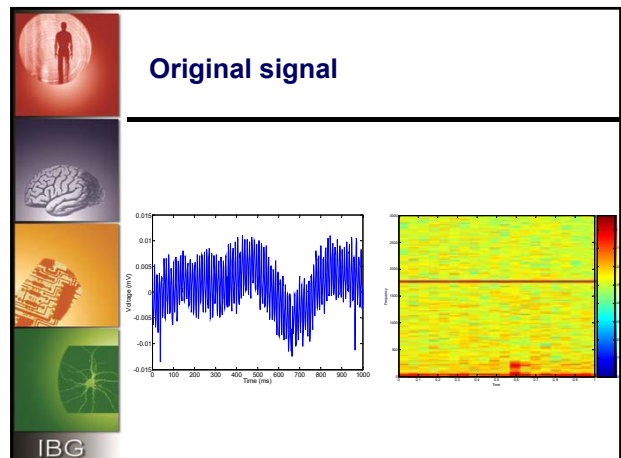
- The absolute power depends heavily on the normalization of the signal.
- The relative power enables detecting the statistics of the signal at unfiltered frequencies.

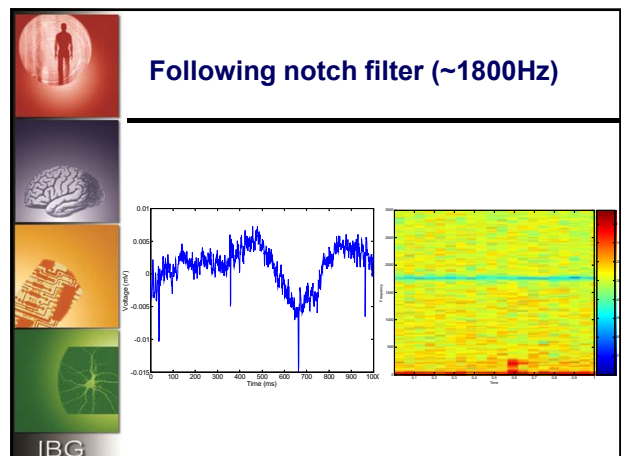
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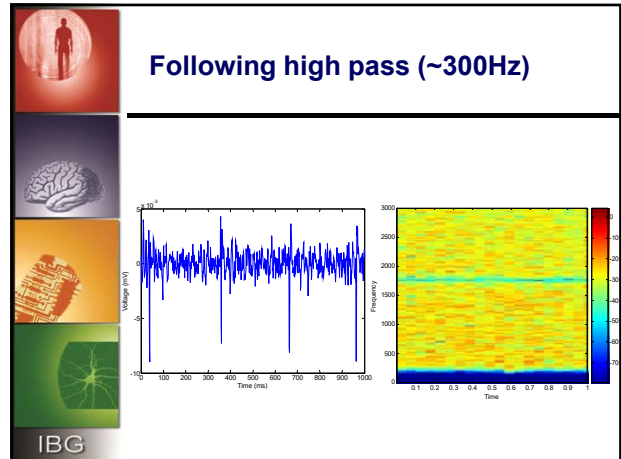

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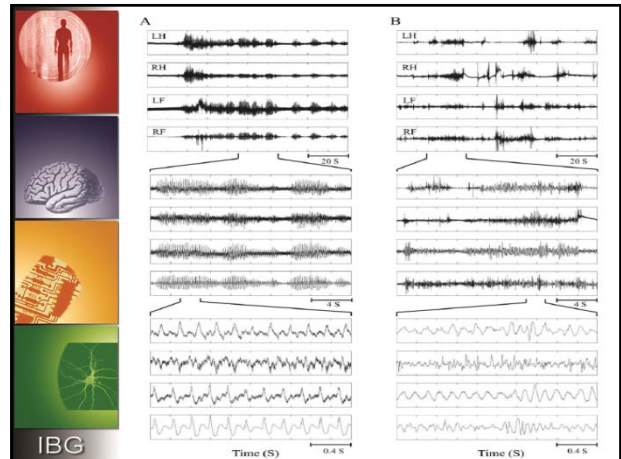
Spectrogram

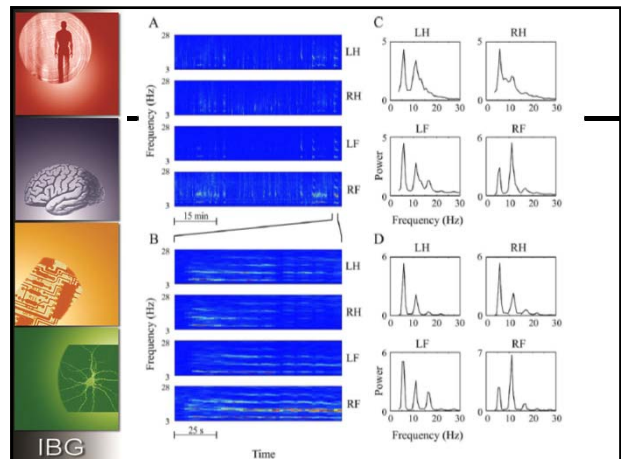
- In many cases the signal is not stationary.
- However, assuming stationarity over short intervals leads to usage of the spectrogram.
- Power spectral density over short periods of time using a sliding window over the signal
- Temporal resolution vs. spectral resolution...

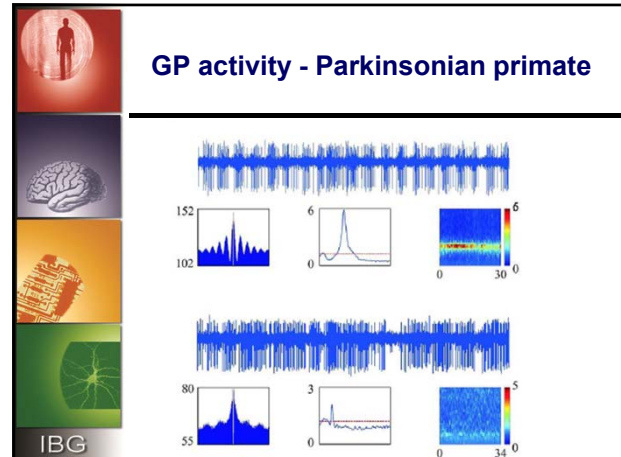


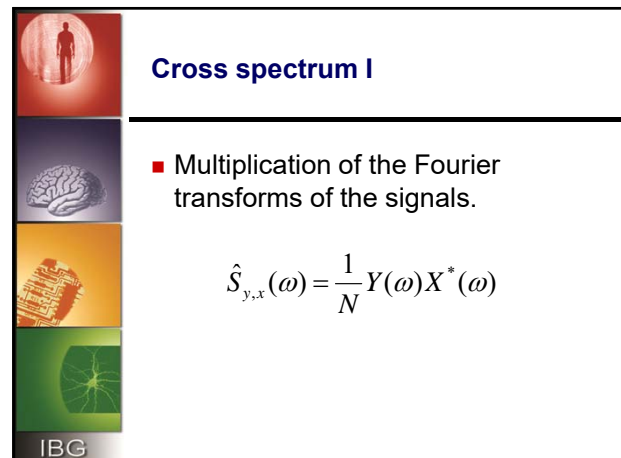


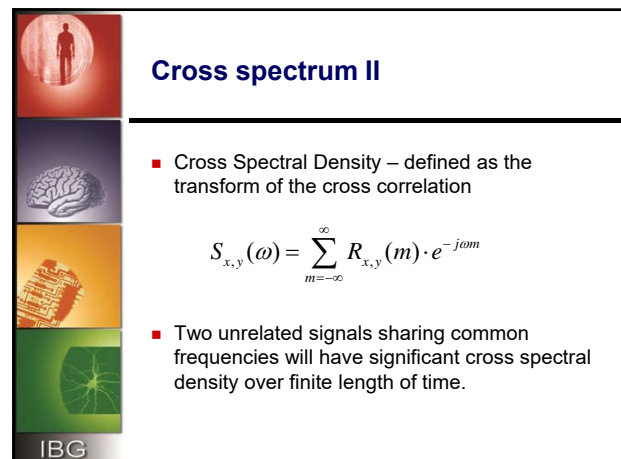


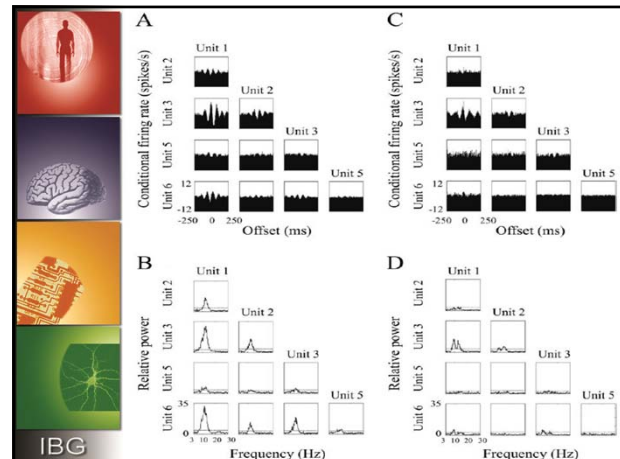












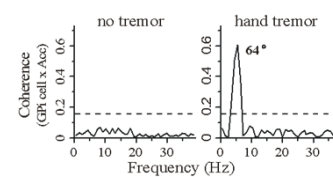
Coherence

- Coherence - ratio of the cross spectral density to the power spectral density of the two signals

$$C_{x,y}(f) = \frac{|S_{x,y}(f)|^2}{S_{x,x}(f) \cdot S_{y,y}(f)}$$

- Normalizes to the spectrum of the two signals and thus relates only to the relation between the signals and not to their structure.

Coherence example



(Taken from Dostrovsky et al)

