



Oktatás, kutatás, gyógyítás: 250 éve az egészség szolgálatában

ANALYSING THE SYNCHRONIZATION OF TIME SERIES – I

Correlation, cross-correlation, cross-spectral analysis, coherence, imaginary coherence, (weighted) phase lag index, functional and effective connectivity

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Methodology of the electrophysiological analysis of sleep-wake states

15. 11. 2022.

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ÁLTALÁNOS ORVOSTUDOMÁNYI KAR Magatartástudományi Intézet Pszichofiziológiai és kronobiológiai kutatócsoport

LOOKING BACK

Fourier spectra



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A HARMONIC IN THE DESCARTES COORDINATES





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EXAMPLE FOR SPECTRAL ANALYSIS: BREATHING AND HEART RATE FREQUENCIES IN WAKEFULNESS AND SLEEP



Breathing (top) and heart rate (tachogram, bottom) time series in wakefulness (WAKE) and in deep, slow wave sleep (SLEEP STAGE 4) ...as well as their power spectral densities (PSD)

Van de Borne P et al. Am J Physiol. 1995 Sep;269(3 Pt 2):H952-8.



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LOOKING BACK

WAVELET ANALYSIS



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Latka M et al. J Physiol Pharmacol 2005, 56, Supp 4, 15-20



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CURRENT PROBLEM

- Depicting the synchronization (connection) between time series (eg. breathing and heart rate)
 - → Is there any synchronization?
 - → How strong it is?
 - → What is its characteristic frequency?
- Sexamples
 - → Activity time series of co-sleeping mother and child
 - Relationship between mood states and timedependent physiological processes
 - Synchronization between distinct physiological signals



A FEW BASIC CONCEPTS

Sconnectivity (coupling) → Functional → Effective Structural Synchronization → Linear ⇒non-linear → mixed



PEARSON CORRELATION

$$r_{xy} = rac{\sum x_i y_i - nar{x}ar{y}}{\sqrt{(\sum x_i^2 - nar{x}^2)}} \sqrt{(\sum y_i^2 - nar{y}^2)}.$$

WHY EXACTLY?

Correlation = Co-variation.

Then what is the question? Are there any more questions?



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PEARSON CORRELATION OF SINE WAVES WITH DIFFERENT AMPLITUDES, BUT IDENTICAL FREQUENCIES AND PHASES (=1)



Series – I.

Senior Research Fellow

... irrespective of their actual frequency





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TIME SERIES MIGHT NOT CORRELATE WITH EACH OTHER BUT THEIR HARMONICS STILL DO



Correlation spectrum



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EEG CORRELATION SPECTRA IN DIFFERENT SLEEP-WAKE STATES





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WHERE DOES CORRELATION WENT? IT DISSAPEARED WITH A TIME DELAY



THE CORRELATION OF TWO TIME SERIES

- And is not constructed to distinguish between frequencies (eg. which of the harmonics instigates the correlation)
- → It can be completed by correlation between FFT (DFT) harmonics (it can be transformed to a frequency-sensitive measure → EEG correlation spectrum)
- It is amplitude free (amplitudes do not matter)!
 It is strongly phase-dependent (brokes down with delay)

Only O-phase differences are well-tolerated



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CROSSCORRELATION = CONVOLUTION Convolving one time series with the another



https://www.purdue.edu/freeform/ervibrations/chapter-iv-animations/convolution-integral-interpreting-resonance/

Similarities will emerge at diff. moments and with diff. frequencies



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CROSSCORRELATION = CONVOLUTION Convolving one of the series with the another





Similarities will emerge at diff. moments and with diff. frequencies



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ABOUT CROSSCORRELATION

Convolution-based method
 It might depict phase delay and common periodicity of the time series





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ON THE WAY TOWARD COHERENCE-I: MULTIPLYING COMPLEX NUMBERS





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ON THE WAY TOWARD COHERENCE-II: THE COMPLEX CONJUGATE





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CROSS-SPECTRAL DENSITY - I

 $a_1 + b_1 i$ Let's take the vector multiplication of one time series with the complex conjugate of the another time series at each harmonic. $a_2 - b_2 i$

$$(a_1 + b_1i)(a_2 - b_2i) = a_1a_2 - a_1b_2i + a_2b_1i - b_1b_2i^2 = (a_1a_2 + b_1b_2) + (a_2b_1 - a_1b_2)i$$

The length of the new vector will be equal with the multiplied length of the two vectors, whereas the phase will be the difference of the two phases (because one was the complex conjugate).

W2: Grin(1024, .01, 3.0, .45)	3: csd(w1,w2)
15	20. U.UU1228 -1.158505d
	or- 26: 0.001720 -1.119901a
***************************************	27: 0.002559 -1.100982a
	28: 0.004171 -1.082528a
05-	29: 0.007894 -1.064523a
	⁰⁴⁻ 30 : 0.019964 -1.046951a
	••• 31 : 0.115169 -1.029799a
45-	32: 0.769464 -1.013055a
	33: 0.037190 -0.996706a
	4
	o
-15 0 1 2 3 4 5 6 7 8 9 10	41
	t b 20 25 30 35 v v b 20 25 30 35 v v v v v v v v v v v v v v v v v v



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CROSS SPECTRAL DENSITY – II



André M. Bastos and Jan-Mathijs Schoffelen Front Syst Neurosci. 2015; 9: 175.



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THE CROSS SPECTRAL DESNITY OF TWO TIME SERIES

- Conveys frequency-specific information
 - It can distinguish between frequencies (which of the

frequencies is decisive in synchronization)

- Amplitude-dependent measure (high input amplitudes
 => apparently high synchronization, ③)
- It is a measure of phase differences
 - It not only signals the 0 phase synchronization and it explicitly tells us about phase differences (in П values or grades) – we can derive a Phase-Slope Index



ONCE AGAIN ABOUT THE ISSUE OF **AMPLITUDES:** let's normalize crossspectral density!

Amplitude of CROSS SPECTRAL DENSITY

The square root of the multiplied AUTOSPECTRAL DENSITY

The numerator and the denominator has to be averaged separately!!!

It can only be interpreted as a mean derived

from multiple pairs of time series

- $coh_{xy}\left(\omega
 ight)=rac{\left|S_{xy}\left(\omega
 ight)
 ight|}{\sqrt{S_{xx}\left(\omega
 ight)S_{yy}\left(\omega
 ight)}}$ Values are between 0 and 1 (from no coherence to maximal coherence)
- Frequency-specific information is conveyed
- Phase differences are not directly measured only phase consistency is reflected



COHERENCE =

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EXAMPLE-I: BREATHING AND HEART RATE TIME SERIES IN WAKEFULNESS AND IN SLEEP



Breathing (top) and heart rate (tachogram, bottom) time series in wakefulness and in deep slow wave sleep

...as well as their coherence

Van de Borne P et al. Am J Physiol. 1995 Sep;269(3 Pt 2):H952-8.



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EXAMPLE-II: INTERHEMISPHERIC COHERENCE OF ALPHA RANGE EEG WAVES CORRELATES WITH THE VOLUME OF THE CALLOSAL REGION (in Alzheimer's disease)





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EXAMPLE-III: DECREASED INTERHEMISPHERIC EEG-COHERENCE IN CORPUS CALLOSUM AGENESIS SYNDROME



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Problems with EEG coherence: active common reference

 EEG always reflects a ptential difference between two recording locations



• Monopolar reference:

coherence will reflect in

part the commonalities

cuased by the activity on

the reference electrode



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Problems with EEG coherence: volume conduction



- Volume conduction: a source spreads to multiple recording locations with the speed of light
- That is, coherence reflects in part the volume conduction
- It is difficult to discern the amount of the separate sources in the signal.



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THEN LET'S DISREGARD THE REAL AXIS (X) : IMAGINARY COHERENCE

- In the formula of coherence function we use the imaginary part of the vector (instead of vector length)
- 0 and 180 grade phase differences is just eliminated from coherence
- As a consequence, there is no volume conduction and common reference problem
- Unfortunately, this method is strongly biased toward 90 grades phase differences, indeed it still considers real axis values in the denominator



(WEIGHTED) PHASE LAG INDEX – (W)PLI



Cornelius J. Stam



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Vinck M et al. NeuroImage 2011;55(4):1548-1565

PHASE LAG INDEXES

PLI

- The <u>modulus</u> of the <u>sign</u> function of cross spectral density
- In case of the values
 scattered <u>around the real</u>
 <u>axis</u> result is close to 0
- It is in fact based on a <u>distribution</u>

WPLI

- Correction of the formula of coherence function with <u>weights derived from the</u> <u>imaginary axis</u>
- If vectors are <u>close to the</u> <u>real axis</u>, their imaginary parts are low
- It is the normalized mean of <u>weighted vectorial</u>
 <u>multiplications</u>



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WPLI IN USE:

Sex differences and

age effects in sleep-

EEG functional

connectivity

Ujma PP et al. Neurobiology of Aging 2019;78:87-97.



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DIRECTION IN FUNCTIONAL CONNECTIVITY-I: Granger-causality

In what measure can a time series be predicted based on its own past (simple linear regression)?

And if we consider the past of another time series as well (multiple linear

regression)? Will the prediction improve significantly?

No - there is no directed connectivity

Yes – there is directed connectivity (we can foresee the future of a time series

based on the past of another)





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DIRECTION IN FUNCTIONAL CONNECTIVITY-II: Phase-Slope Index (PSI)

- Interactions are time consuming
- If propagation speed is equal for different frequency oscillations
- Then: phase differences between the transmitter and the receiver increase in a frequency-dependent manner

Slope of the phase angle derived from the cross spectral density

Nolte G et al. Phys Rev Lett 2008



NREM sleep: right central EEG and right anterior thalamic LFP coherence: increased values in the slpindle range

Calculating the PSI from the phase angles of the cross spectral density

Tsai et al. Neurosci Lett 2010



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WHAT ARE THE PROBLEMS WITH LINEAR METHODS AND UNIDIRECTIONALITY?

- Non-linear interactions (eg. Large increases above a certain threshold) result in misrepresentations in linear connectivity
- Can we really exclude unidirectionality in neural interactions?
 - →Just very rarely and hypothetically (see for example the case of thalamo-cortico-thalamic interactions)



MUTUAL SYMBOLIC INFORMATION



It can complete WPLI-measurements

<u>Sci Rep</u>. 2019; 9: 8894.



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CLOSING REMARKS

♥ We could still discuss:

Instantaneous phase differences – Hilbert transformation (perhaps during the last lecture)

♥ Questions? ☺



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