

Non-Parametric Rank Statistics for Spectral Power and Coherence

Nasseroleslami B*, Dukic S, Bista S, Buxo T, Coffey A, McMackin R, Muthuraman M, Hardiman O, Lalor EC*, Lowery MM*

Bahman Nasseroleslami, BSc, MSc, PhD

Fr Tony Coote Assistant Professor in Neuroelectric Signal Analysis in MND Principal Investigator and Research Strand Leader, Academic Unit of Neurology, School of Medicine Trinity College Dublin, the University of Dublin Dublin, Ireland

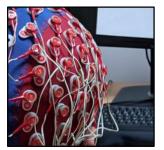
Wednesday 02/12/2020 School of Computer Science and Statistics Weekly Seminars, Trinity College Dublin, the University of Dublin

- 0. Motivation (Application Domain)
- 1. Introduction
- 2. Original Formulation for Spectra and Coherence
- 3. Non-Parametric Estimation of Spectra and Coherence
- 4. No-Parametric Rank Statistics for Spectral Power
 - 1-Sample Power
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 - 1. 1-Sample Coherence
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Motivation

Targeting and Interrogating Networks by Novel Neuro-electric Biomarkers



EEG (with source Analysis)



EMG (multi-channel/HD)



TMS (+ Threshold Tracking)

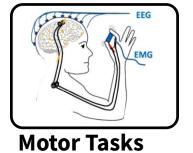
Paradigms

Motivation

Targeting and Interrogating Networks by Novel Neuro-electric Biomarkers

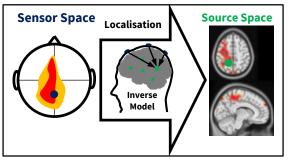




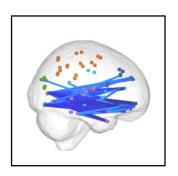


Motivation

Targeting and Interrogating Networks by Novel Neuro-electric Biomarkers



Brain Source Reconstruction

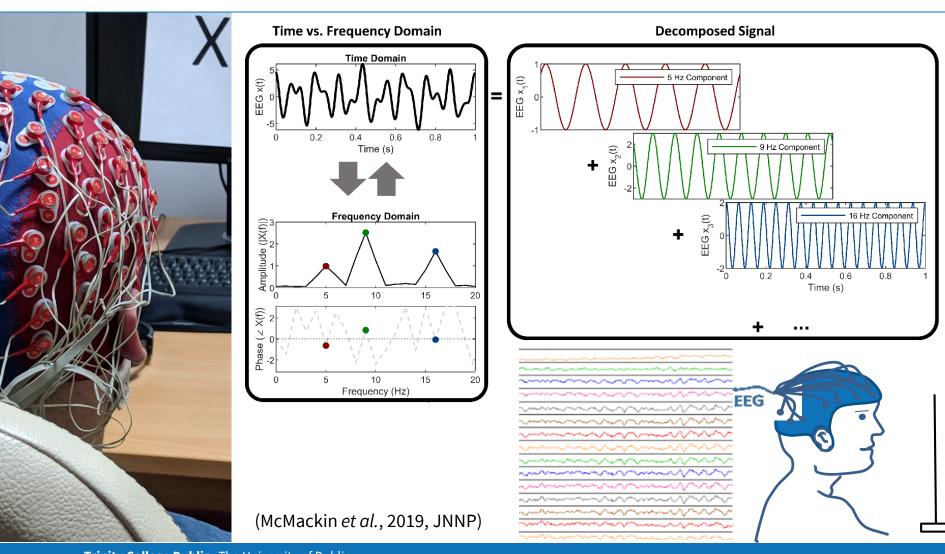


(McMackin *et al.*, 2019, J Neurol Neuros Psych)

Connectivity Analysis

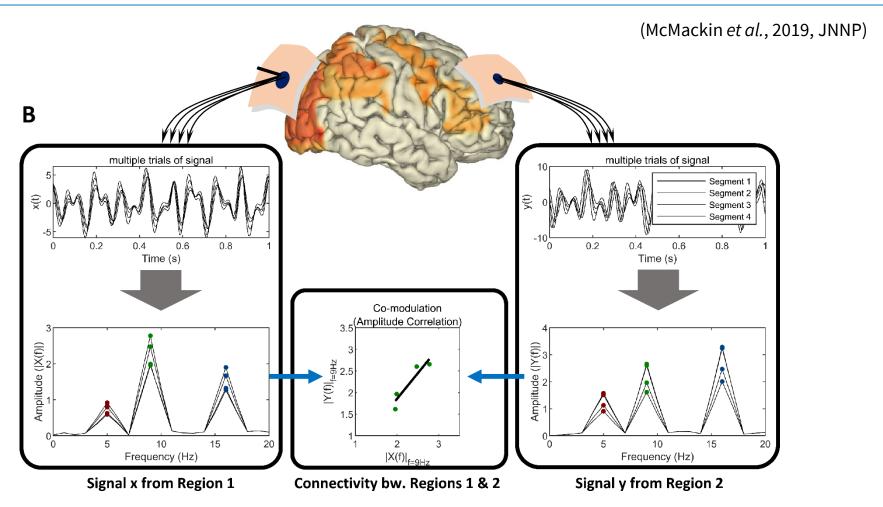
Resting-State EEG Networks

Resting State: Continuous EEG Recordings



Motivation: Resting-State EEG Networks

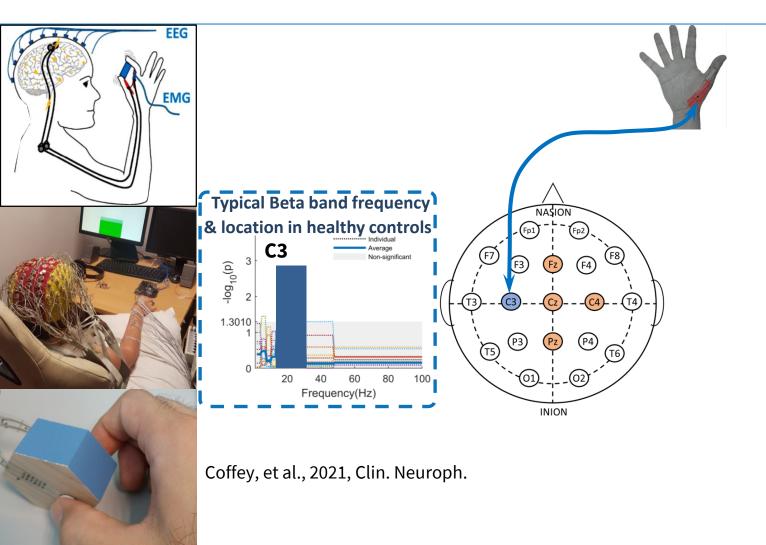
Spectral Power and Connectivity (Co-Modulation and/or Synchrony)



Frequency-Specific (Spectral) Analysis of Power and Connectivity/Synchrony

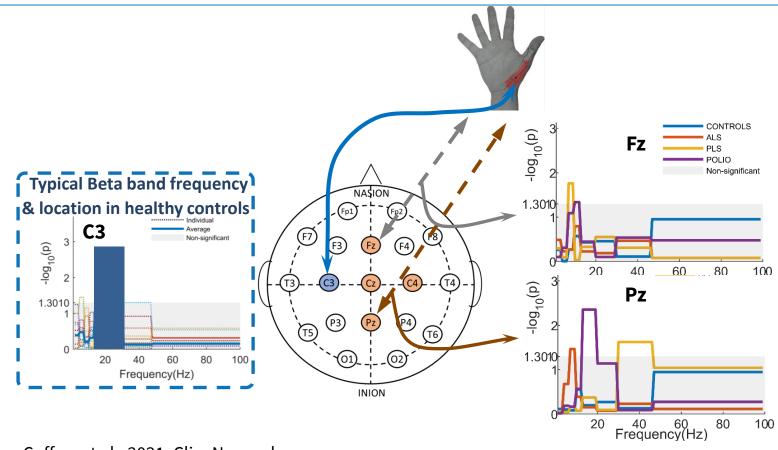
Motivation: Motor Networks

Motor Tasks: EEG-EMG Coherence



Motivation: Motor Networks

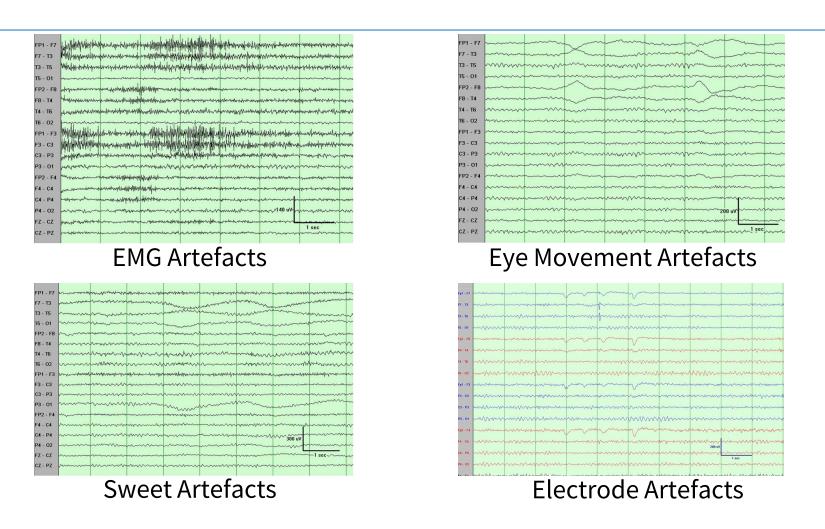
Increase Abnormal EEG-EMG Coherence in MND/ALS



Coffey, et al., 2021, Clin. Neuroph.

Non-motor locations & abnormal frequencies in PLS,POL, ALS show compensatory activity in other cortico-spinal networks

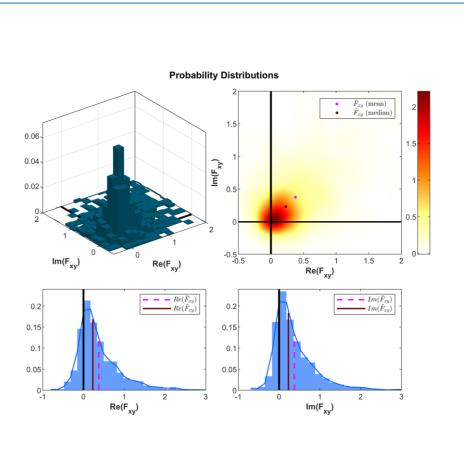
Motivation Examples of EEG vs Artefacts



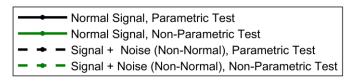
http://emedicine.medscape.com/article/1140247-overview

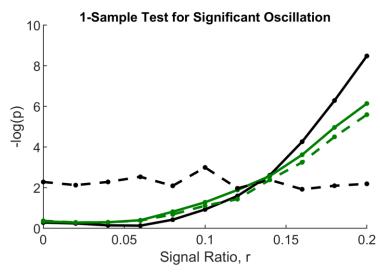
Overview

Non-Parametric Rank Statistics for Spectral Power and Coherence



Robustness of Non-Parametric Tests for Simulated Spectral Power





Nasseroleslami et al., BioRxiv

Non-Parametric-based Estimates of Power and Coherence are Robust against Artefacts

Dukic et al., 2017, IEEE EMBC 2017

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Introduction

- Parametric Estimation of Spectral Power and Coherence in (Neural) Time Series Analysis is documented (Brillinger, 2001; Halliday and Rosenberg, 1999)
- The statistical inference of spectral power and coherence of neural signals remains a practical challenge.
 - Non-normal Distribution
 - Artefactual Components
 - Bias
 - Complex statistical distributions.

Introduction

• To parallel, Sign rank, Mann-Whitney tests, we need nonparametric methods.

• Non-parametric methods (e.g. based on median) of spectra afford robust estimation (Dukic et al. 2017).

Estimation of Coherence using the Median is Robust against EEG Artefacts

Stefan Dukic, Parameswaran Mahadeva Iyer, Kieran Mohr, Orla Hardiman, Edmund C. Lalor, Bahman Nasseroleslami

 However, the statistical inference based on these non-parametric estimates remain to be formulated and tested.

Aim and Objective:

 Aim: To provide non-parametric rank tests for 1- and 2-sample statistical testing of spectral power and coherence

• To demonstrate and verify the non-parametric tests using simulated and real neural signals in different conditions, and to assess their robustness in presence of artefactual components.

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Spectral Power and Coherence

Spectral Power

Consider x(t) and y(t) to be time domain signals

$$F_{xx}(f) = \mathcal{E}\{X_i(f)X_i(f)^*\}$$

$$F_{yy}(f) = \mathcal{E}\{Y_i(f)Y_i(f)^*\}$$

$$F_{xy}(f) = \mathcal{E}\{X_i(f)Y_i(f)^*\}$$

$$\bar{F}_{xx}(f) = \frac{1}{L} \sum_{i=1}^{L} X_i(f) X_i(f)^*$$

$$\bar{F}_{yy}(f) = \frac{1}{L} \sum_{i=1}^{L} Y_i(f) Y_i(f)^*$$

$$\bar{F}_{xy}(f) = \frac{1}{L} \sum_{i=1}^{L} X_i(f) Y_i(f)^*$$

Statistics:

$$var\{F_{xx}(f)\} \approx (F_{xx}(f))^2/L$$

Probability Distributions 0.1 0.2 0.05 0.1 -2 0 4 **آ** 4 x(t) 0.3 0.1 0.2 0.05 0.1 -2 0 2 4 2 4 $\mathsf{F}_{\mathsf{y}\mathsf{y}}$ y(t)

Spectral Power and Coherence

Coherence

• Coherency function C and coherence $|C_{xy}(f)|^2$

$$C_{xy}(f) = \frac{F_{xy}(f)}{\sqrt{F_{xx}(f)F_{yy}(f)}}$$

- Statistics:
- Coherence has a hypergeometric (sampling) distribution. Under null hypothesis (0 coherence), $tanh^{-1}(.)$ provides an approximate transformation to normal.

$$p = (1 - |C_{xy}(f)|^2)^{(L-1)}$$

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Spectral Power

Consider x(t) and y(t) to be time domain signals

$$F_{xx}(f) = \mathcal{E}\{X_i(f)X_i(f)^*\}$$

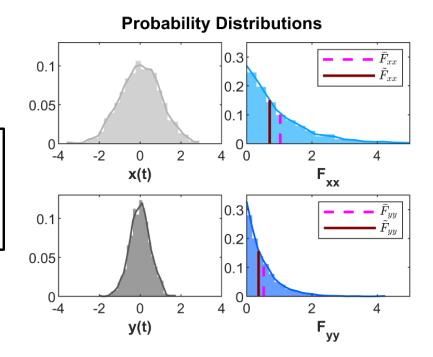
$$F_{yy}(f) = \mathcal{E}\{Y_i(f)Y_i(f)^*\}$$

$$F_{xy}(f) = \mathcal{E}\{X_i(f)Y_i(f)^*\}$$

$$\tilde{F}_{xx}(f) = Median_i[\{X_i(f)X_i(f)^*\}]$$

$$\tilde{F}_{yy}(f) = Median_i[\{Y_i(f)Y_i(f)^*\}]$$

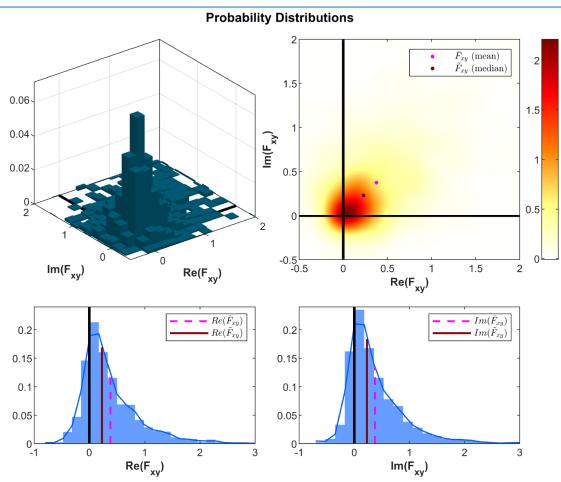
$$\tilde{F}_{xy}(f) = Median_i[\{X_i(f)Y_i(f)^*\}]$$



Spatial Median:

$$\tilde{F}_{xy}(f) = \underset{\Theta}{\operatorname{arg\,min}} (\sum_{i=1}^{L} \|X_i(f)Y_i(f)^* - \Theta\|)$$

Coherence



Statistical Distribution of the Raw Cross-Spectra in 2D and Marginal Plots and the Estimation of Cross-Spectrum using the Mean and Median (\sim Fxy).

Coherence

Consider x(t) and y(t) to be time domain signals

$$F_{xx}(f) = \mathcal{E}\{X_i(f)X_i(f)^*\}$$

$$F_{yy}(f) = \mathcal{E}\{Y_i(f)Y_i(f)^*\}$$

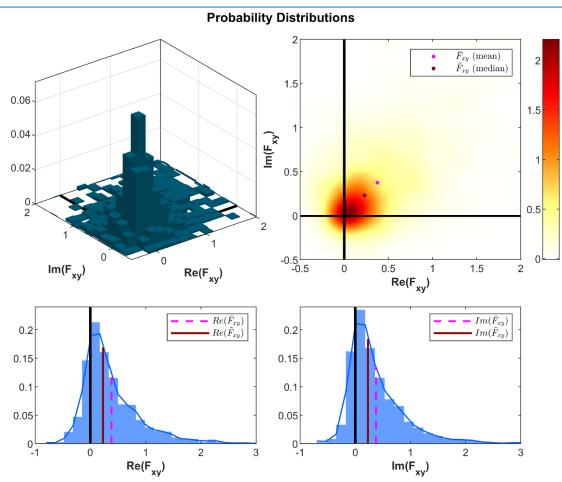
$$F_{xy}(f) = \mathcal{E}\{X_i(f)Y_i(f)^*\}$$

Spatial Median:

$$\tilde{F}_{xy}(f) = \underset{\Theta}{\operatorname{arg\,min}} (\sum_{i=1}^{L} \|X_i(f)Y_i(f)^* - \Theta\|)$$

$$\tilde{C}_{xy}(f) = \frac{\tilde{F}_{xy}(f)}{\sqrt{\tilde{F}_{xx}(f)\tilde{F}_{yy}(f)}}$$

Coherence



Statistical Distribution of the Raw Cross-Spectra in 2D and Marginal Plots and the Estimation of Cross-Spectrum using the Mean and Median.

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Non-Parametric Rank Statistics for Spectral Power

Main change in the way we handle time series statistics:

Exploiting Individual and All Data Points Regardless of Their Measures and Their (Sampling) Distributions

Non-Parametric Rank Statistics for Spectral Power

Confidence Intervals from Raw Spectra:

$$[iCDF_{\{|X_i(f)|^2\}}(\alpha/2),iCDF_{\{|X_i(f)|^2\}}(1-(\alpha/2))]$$

Non-Parametric Rank Statistics for Spectral Power

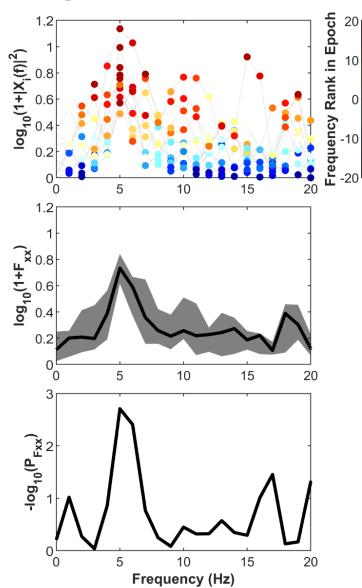
1 Sample

Typical Hypothesis: Presence of significant (decrease/increase) in specific frequencies (compared to white noise)

One-Sample Significant Power: Wilcoxon's Signed Rank test on the centred rank values

2 Sample

Two-Sample Significant Power Difference: Traditional Mann-Whitney U test.



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Non-Parametric Rank Statistics for Coherence

1 Sample

1-Sample Location Problem

Test Selection Criterion:

Affine Invariant

2.5 2.5 $\{X_iY_i\}$ \tilde{F}_{xy} (median) 1.5 Im(F_{xy}) 1.5 0.5 0.5 -0.5 0 2 $Re(F_{xy})$

Significance Testing for Coherence

One-Sample Significant Coherence: One-Sample Spatial (signed) Ranks Test (Hannu, Oja & Randles, 2004; Hannu, Oja, 2010; Nordhausen & Oja, 2011).

Non-Parametric Rank Statistics for Coherence

2 Sample

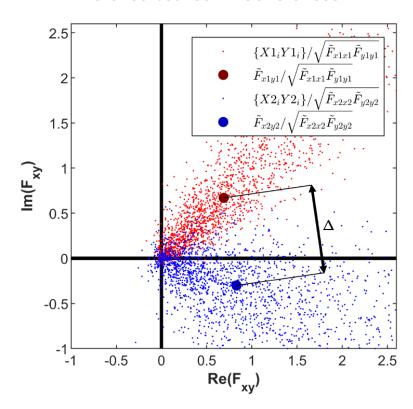
2-Sample Location Problem

Test Selection Criterion:

Affine Invariant

• Reflects Phase or Magnitude Difference

Difference between 2 Coherences



Two-Sample Significant Coherence Difference (Magnitude and Phase): Two-Sample Spatial Ranks Test (Hannu Oja & Randles, 2004; Hannu Oja, 2010; Nordhausen & Oja, 2011).

Non-Parametric Rank Statistics for Coherence

2 Sample: Separate Testing for Phase or Magnitude

Two-Sample Significant Difference in Coherence
 Magnitude: Statistical trick by subtraction of z-scores, similar to Stouffer's method.

$$z_{1} = iCDF_{\mathcal{N}(0,1)}(p_{1})$$

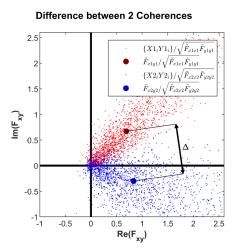
$$z_{2} = iCDF_{\mathcal{N}(0,1)}(p_{2})$$

$$P_{diff,L} = CDF_{\mathcal{N}(0,1)}(\frac{z_{2}-z_{1}}{\sqrt{2}})$$

$$P_{diff,R} = 1 - P_{diff,L}$$

$$P_{diff} = 2.min(P_{diff,L}, P_{diff,R})$$

• Two-Sample Significant Difference in Coherence Phase: Mann-Whitney U Test between the global-meansubtracted phase values. (or other circular statistical tests)



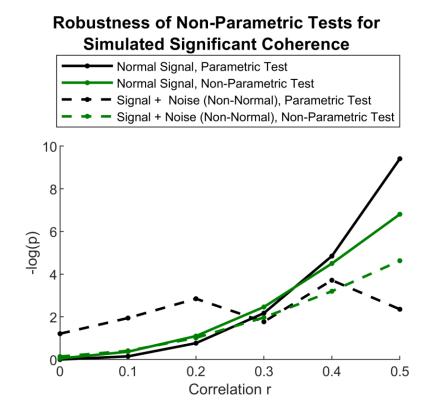
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Numerical Examples

Simulations

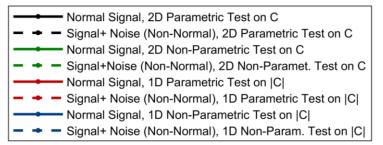


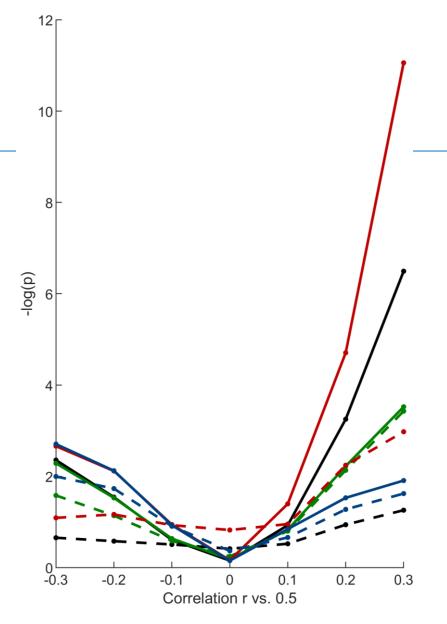
The 1-Sample Spatial Signed Rank, is Robust Against Artefacts.

Numerical Examples

Simulations

Robustness of Non-Parametric Tests for Simulated Difference in Coherence Magnitude



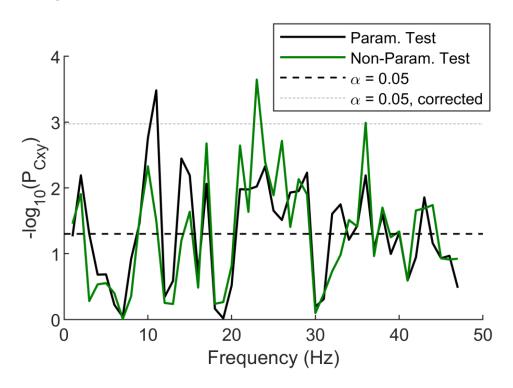


The 2-Sample Statistical Tests based on SpatialRank and Spatial SignedRank are Robust Against Artefacts

Numerical Examples

Real Data

Significance of Experimental Cortico-Muscular Coherence



Both Test Families Detect Significant Presence of Coherence Between EEG and EMG signals.

Figure: Difference Between Cortico-Muscular Coherence in Left/Right Hemispheres.

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Discussion

• **Benefits:** Robustness, Distribution-free, Allowing Testing for Magnitude/Phase, No need for Bootstrapping.

Challenges: Closed form solutions, lower statistical power and sensitivity

• **Opportunities:** Usable for Time-Frequency, and Partial spectral and Coherence, and other time series, can be tested at individual subject or group-level.

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Recommendations

Summary of Tests

- One-Sample Significant Power: Wilcoxon's Signed Rank test on the centred rank values [1].
- Two-Sample Significant Power Difference: Traditional Mann-Whitney U test.
- One-Sample Significant Coherence: One-Sample Spatial (signed) Ranks Test (Hannu, Oja & Randles, 2004; Hannu, Oja, 2010; Nordhausen & Oja, 2011).
- Two-Sample Significant Coherence Difference (Magnitude and Phase): Two-Sample Spatial Ranks Test (Hannu Oja & Randles, 2004; Hannu Oja, 2010; Nordhausen & Oja, 2011).
- Two-Sample Significant Difference in Coherence Magnitude: Statistical trick by subtraction of z-scores, similar to Stouffer's method [1].
- Two-Sample Significant Difference in Coherence Phase: Mann-Whitney U Test between the global-mean-subtracted phase values.

Conclusions

- The approach provides a new framework for non-parametric statistical analysis of the neural signal spectra.
- These methods are suited to neuroscience & neural engineering applications, given the attractive properties such as minimal assumption on distributions, statistical robustness, and the diverse testing scenarios afforded.

Reference

Nasseroleslami B, Dukic S, Bista S, Buxo T, Coffey A, McMackin R, Muthuraman M, Hardiman O, Lalor EC, Lowery MM. 2019. "Non-Parametric Rank Statistics for Spectral Power and Coherence". bioRxiv.:818906. doi:10.1101/818906.

Thank you

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Thank You

