

The Impact of Using Effective Connectivity Measures (Granger Causality) in Guiding Neurofeedback



ISNR

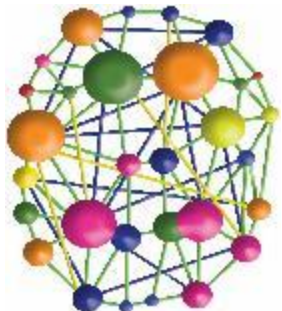
26th Annual Conference

Location: Glendale, Arizona

Conference: October 18–21, 2018

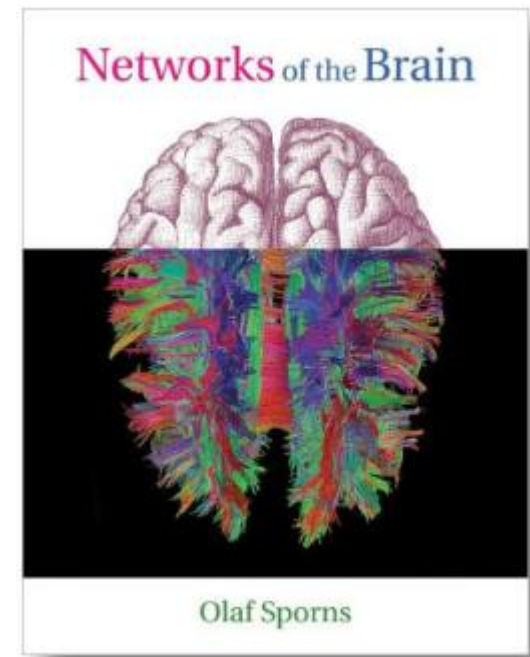
Pre-conference: October 15–17, 2018

Robert Coben, PhD & Anne W Stevens, PhD,
Integrated Neuroscience Services



Types of Neural Connectivity

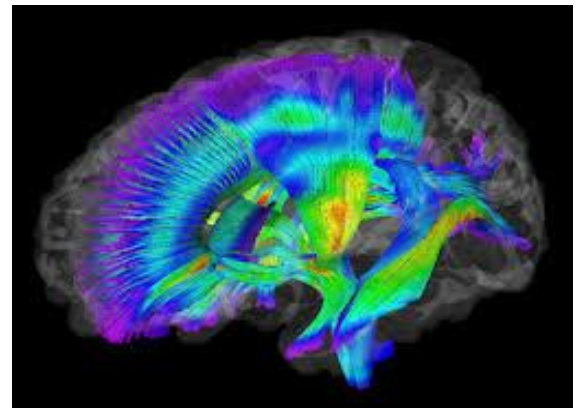
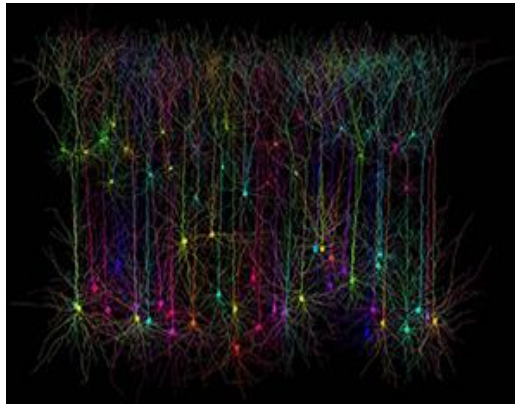
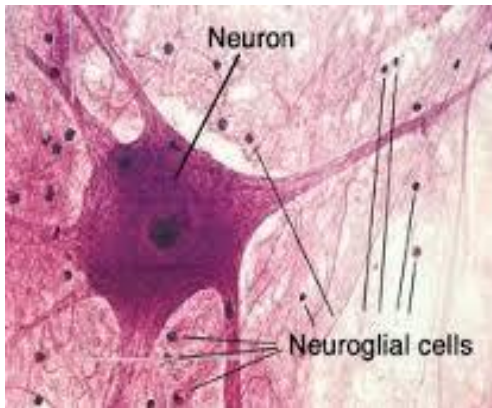
- ◎ ***Structural Connectivity***
- ◎ ***Functional Connectivity***
- ◎ ***Effective Connectivity***



Sporns, O. (2010). *Networks of the Brain*. MIT press.

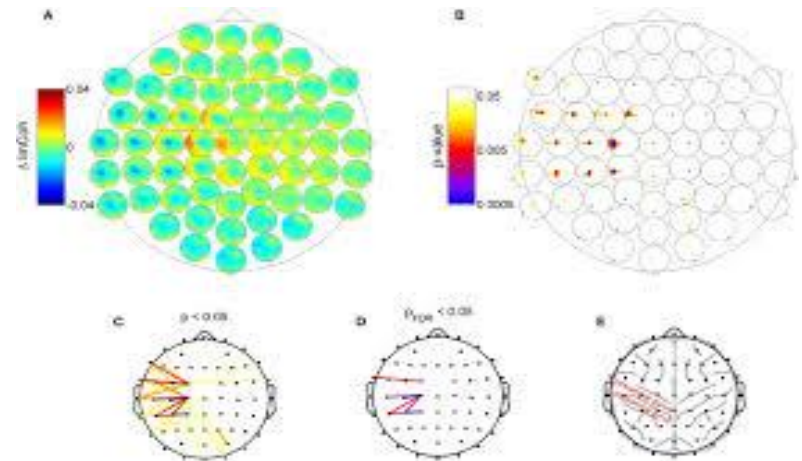
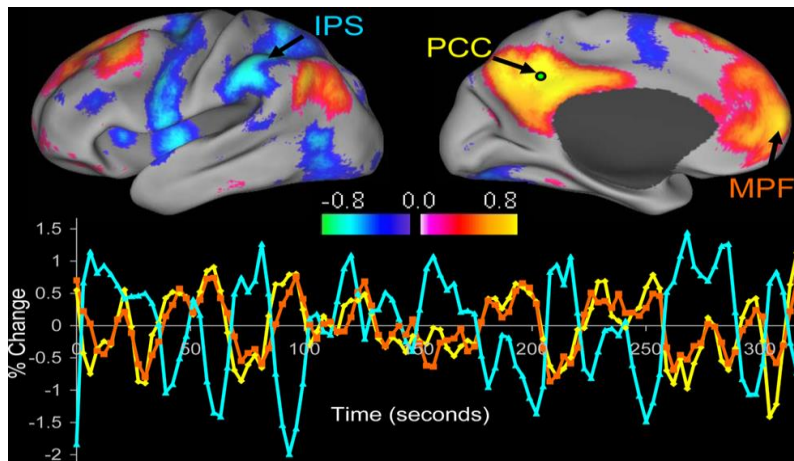
Structural Connectivity

⊙ A set of physical or anatomical connections linking neural elements.



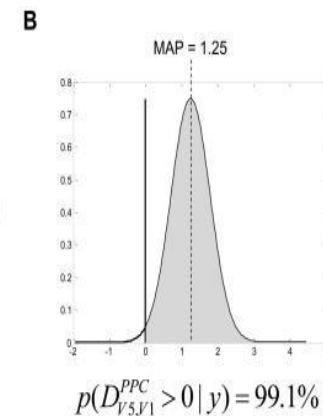
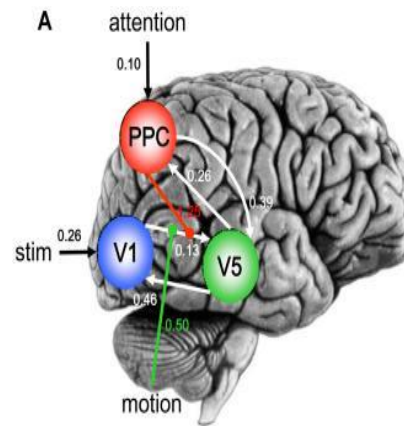
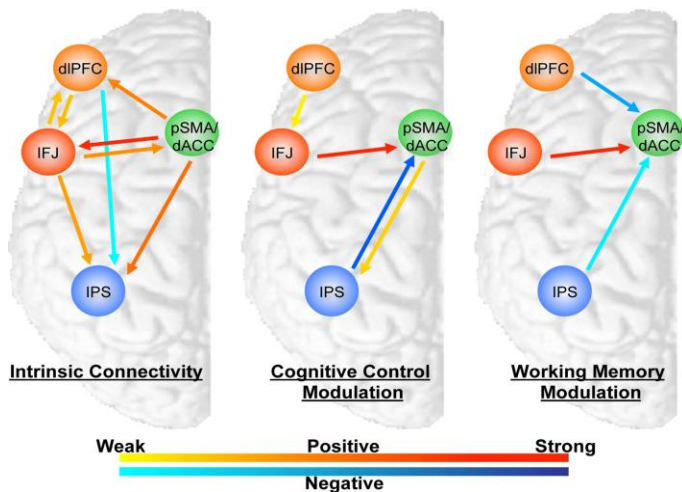
Functional Connectivity

⊙ Patterns of deviations from statistical independence between distributed and spatially remote neuronal units. The basis of this is time series data from neural recordings. Their relation is taken as neuronal coupling and often takes the form of correlation, coherence, phase locking or comodulation. There is no causal relationship, effect or interaction.



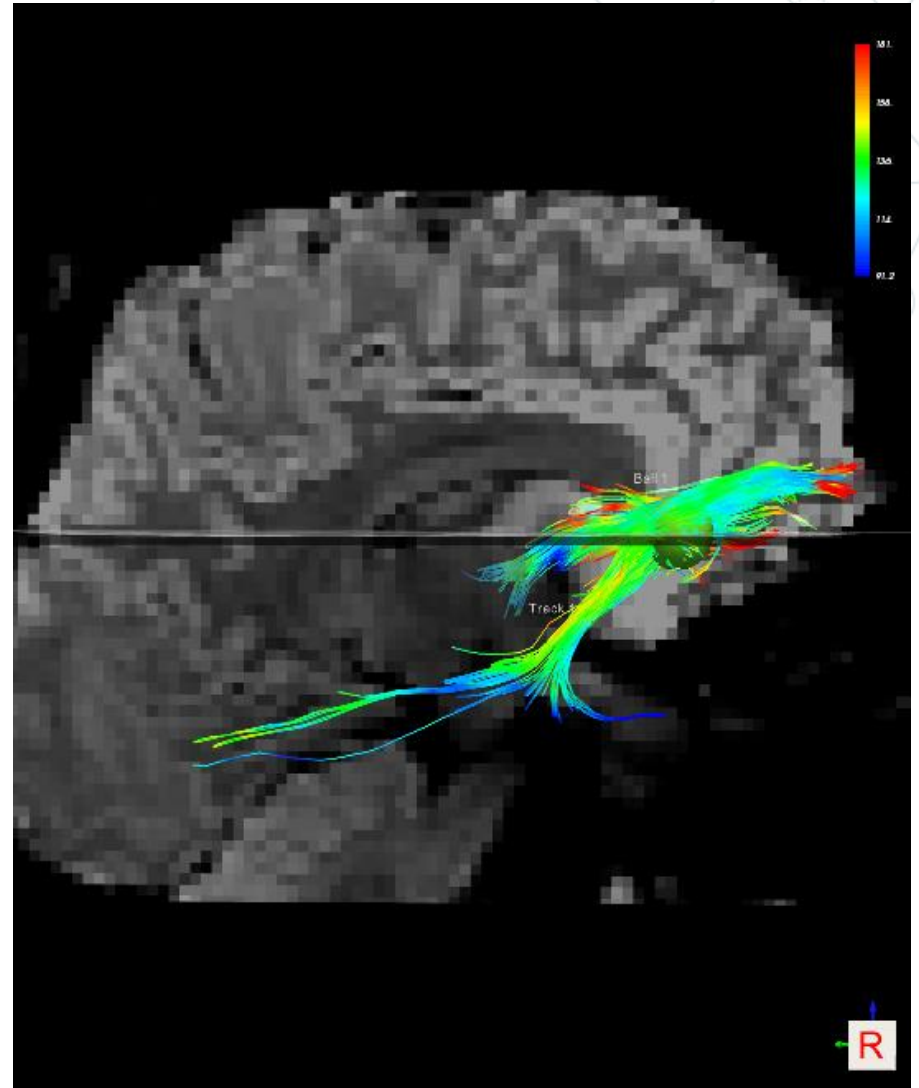
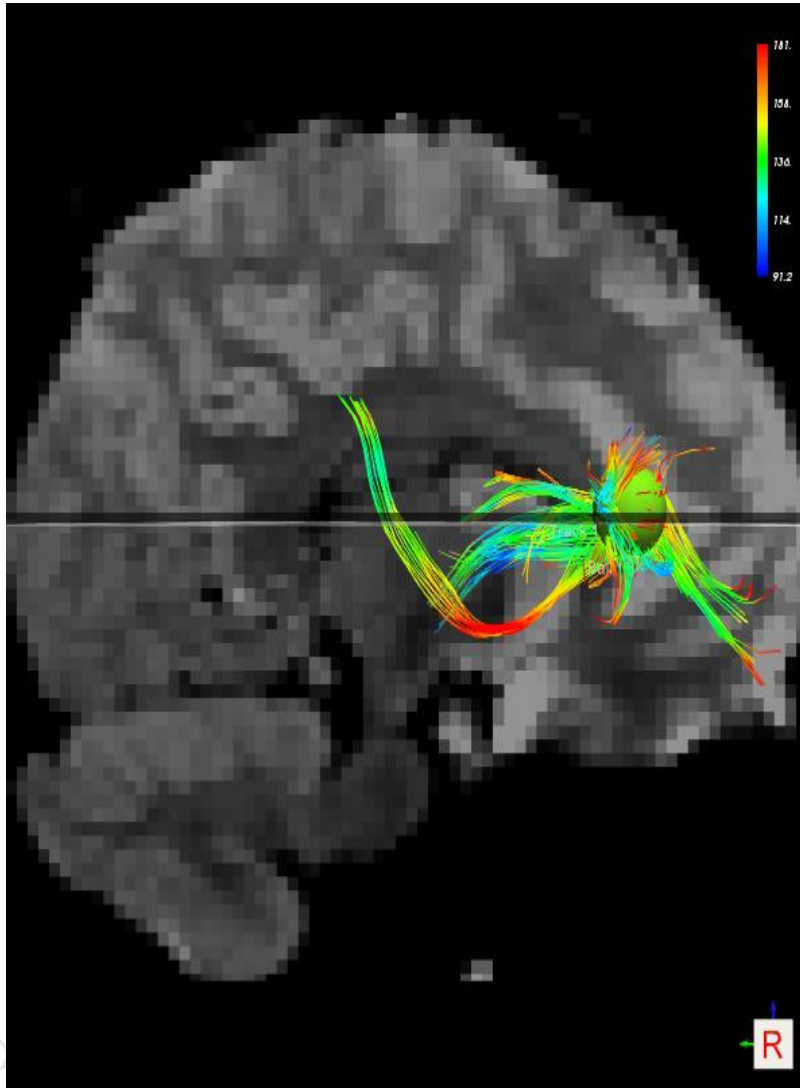
Effective Connectivity

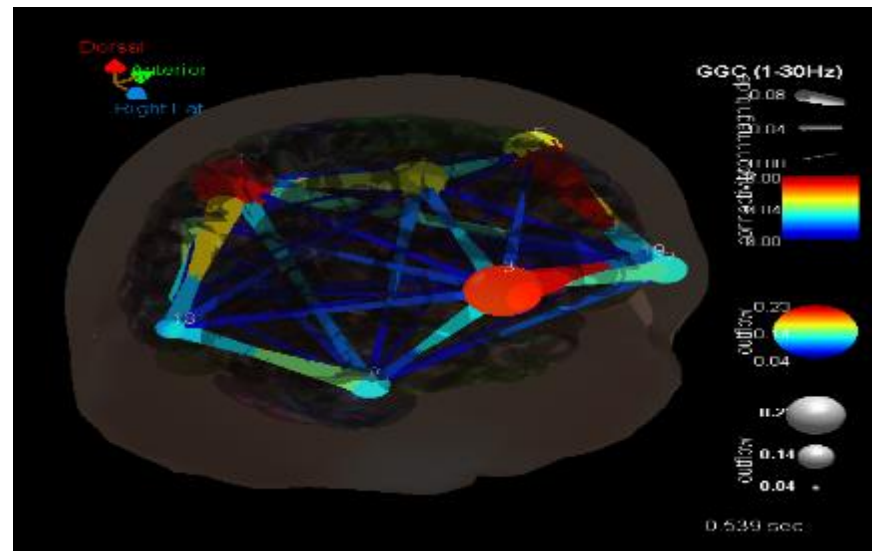
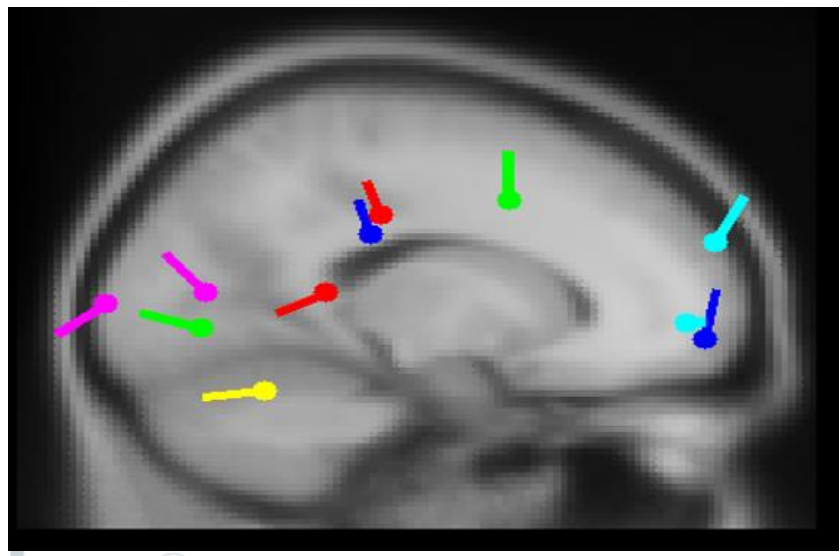
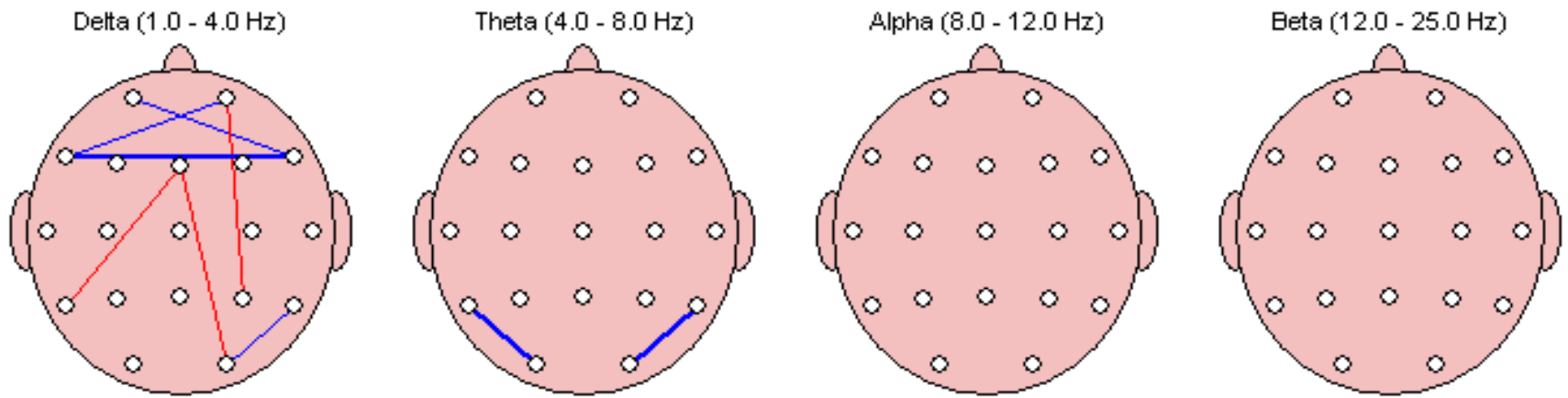
- ⊙ Network and causal effects between neural elements.
- ⊙ Inferred through statistical techniques such as time series analyses and statistical modeling that assess causality and interaction.
- ⊙ Requires complex data processing and modeling techniques such as ICA, Partial Directed Coherence and Granger Causality.



Multivariate
Autoregressive
Models (MVAR)

Comparing levels of connectivity





Review of the methods of determination of directed connectivity from multichannel data

Katarzyna J. Blinowska

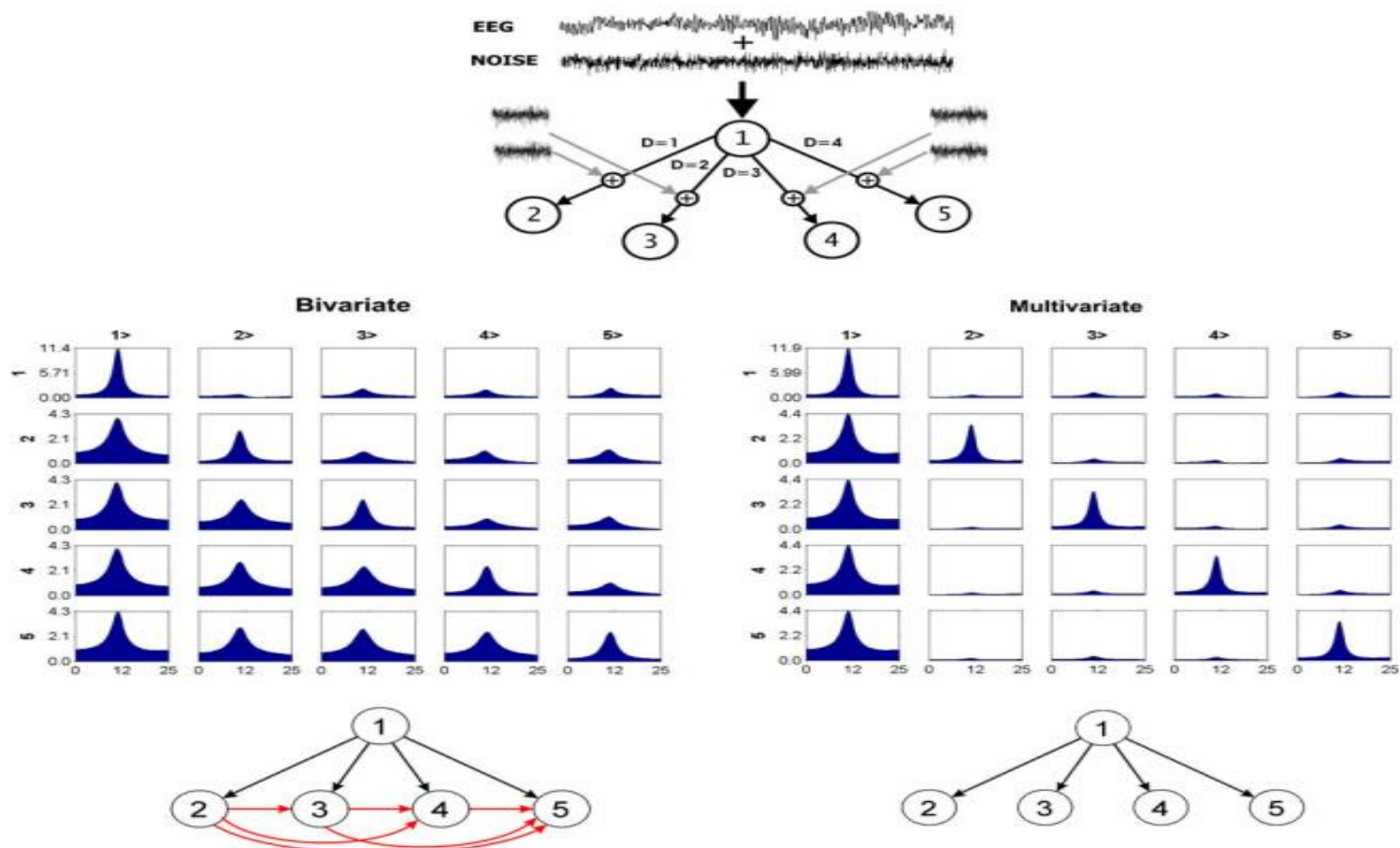


Fig. 1 Comparison of bivariate and multivariate methods of estimation of directed connectivity. *Top* simulation scheme (D delay value, at each step white noise is added). *Bottom* connectivity measures, at the *left* bivariate, at the *right* multivariate. Propagation from the

channel marked above the column to the channel marked at *left*. In each *box* DTF is shown as a function of frequency. At the *diagonal* power spectra. At the very *bottom* obtained connections schemes

Is Graph Theoretical Analysis a Useful Tool for Quantification of Connectivity Obtained by Means of EEG/MEG Techniques?

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¹Department of Biomedical Physics, Faculty of Physics, University of Warsaw, Warsaw, Poland

²Institute of Biocybernetics and Biomedical Engineering of Polish Academy of Sciences, Warsaw, Poland

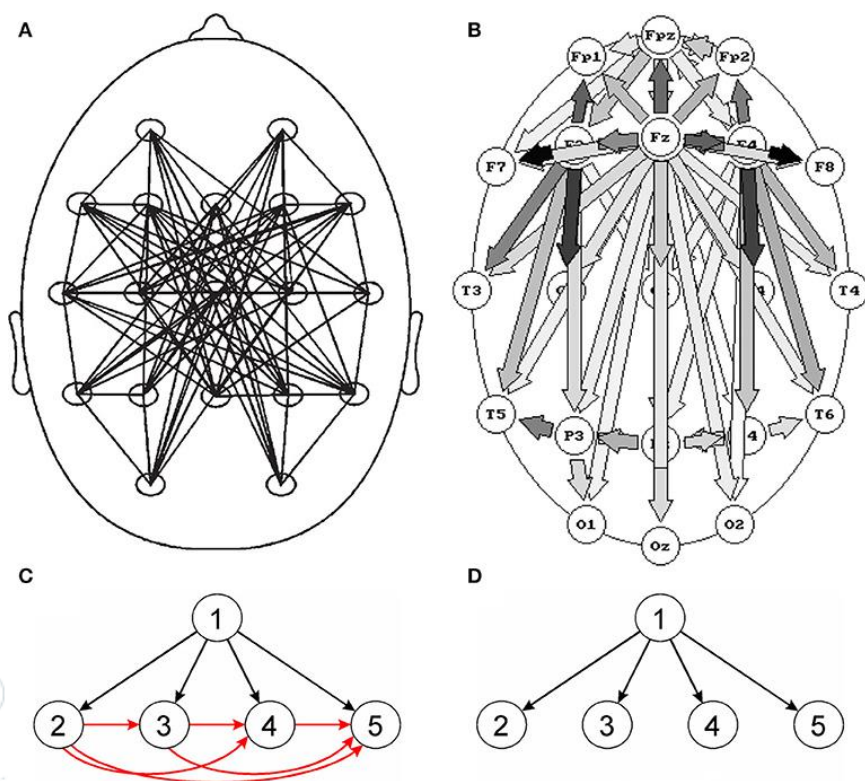


Figure 1. Comparison of bivariate and multivariate connectivity measures. Top images: connectivity patterns for slow wave sleep (stage 3/4), (A) obtained using the bivariate measure (SL), (B) obtained using the multivariate measure (DTF). Although in (A) undirected and in (B) directed connections are shown, however the main difference between the pictures are: disorganized pattern of connections in (A) and clear-cut pattern of connections compatible with physiological evidence in (B). Bottom images – propagation patterns for a simulation which assumes a propagation of activity from electrode 1 to electrodes 2, 3, 4, and 5; (C) – pattern obtained for a bivariate measure (coherence) and (D) – for a multivariate measure (DTF). For the bivariate connectivity measure, false connections are created

resulting from common driving. (A) Reproduced from [Leistedt et al. \(2009\)](#). (B) Reproduced from [Kaminski et al. \(1997\)](#) (with permission).

Using quantitative and analytic EEG methods in the understanding of connectivity in autism spectrum disorders: a theory of mixed over- and under-connectivity

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³ Center for Mind and Brain, University of California, Davis, CA, USA

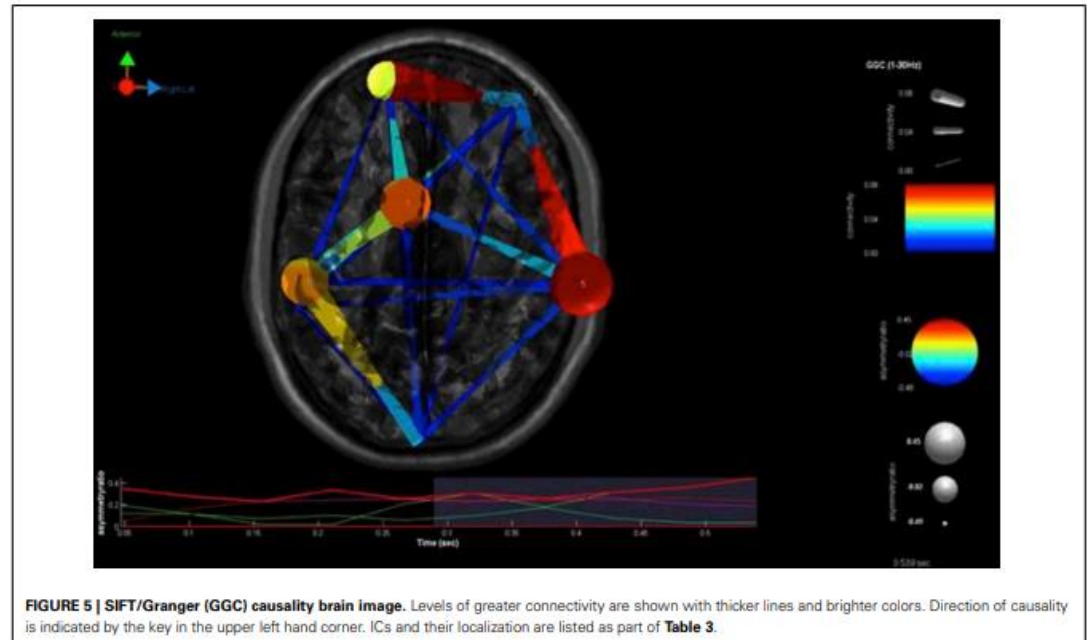
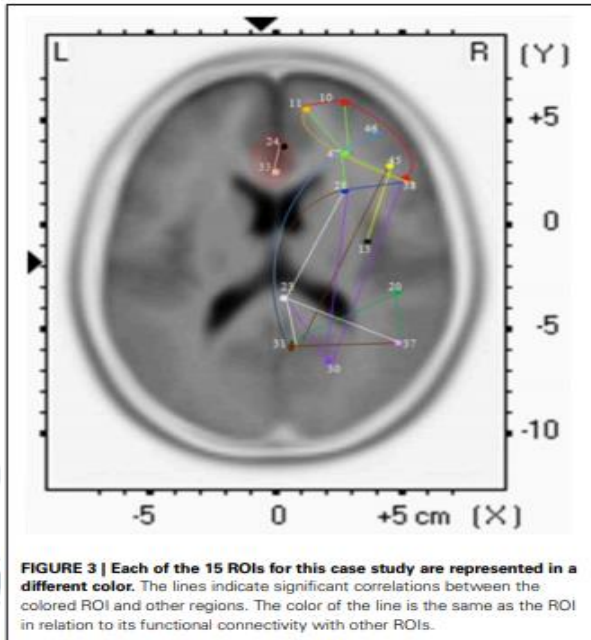
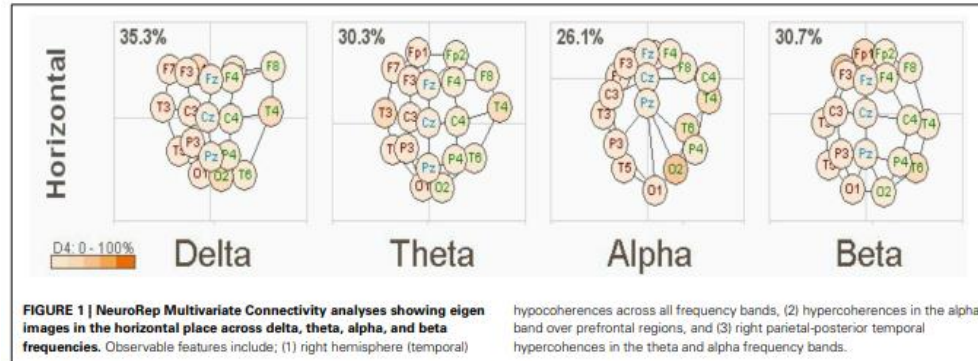
⁴ Semel Institute for Neuroscience and Human Behavior, University of California, Los Angeles, CA, USA

⁵ Psychoeducational Network, Knoxville, TN, USA

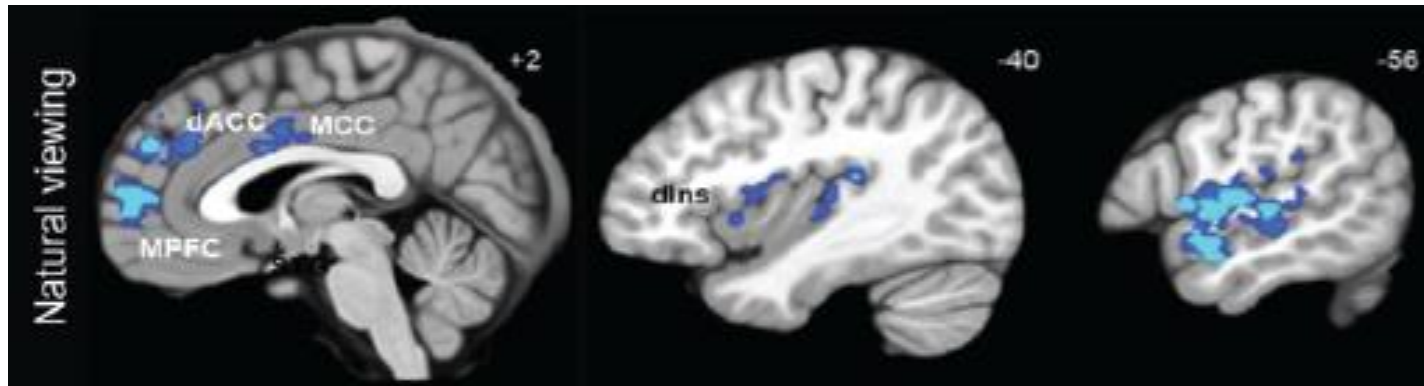
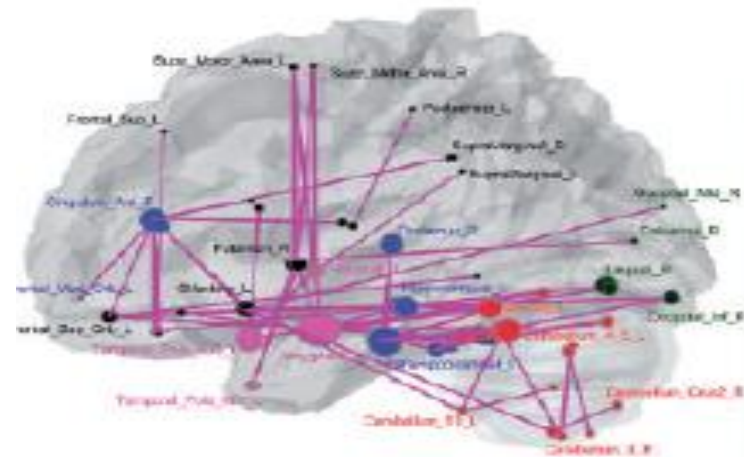
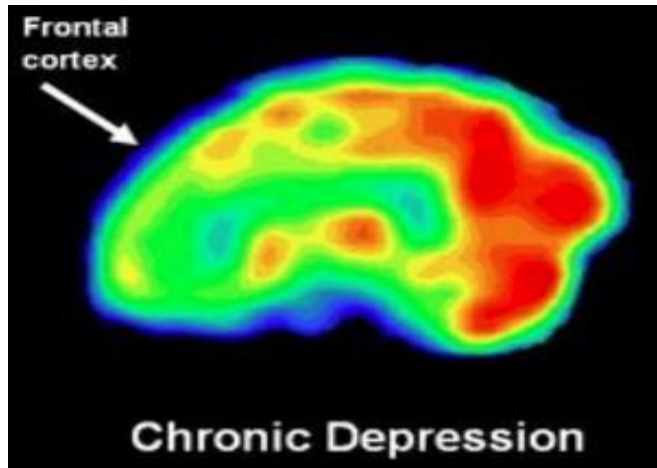
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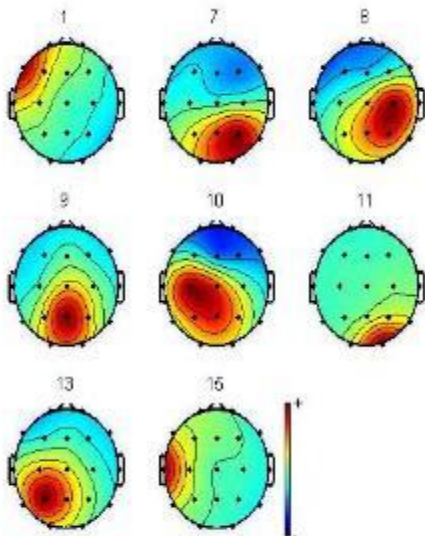
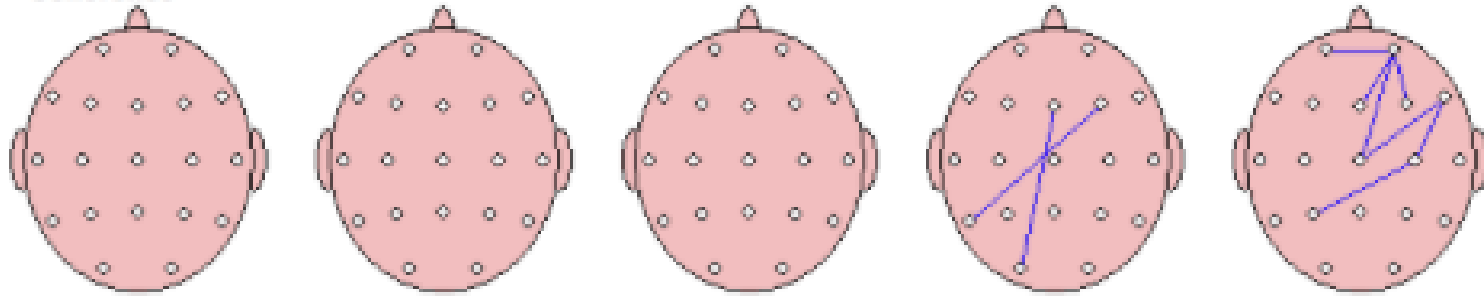


Exemplar: Major Depression

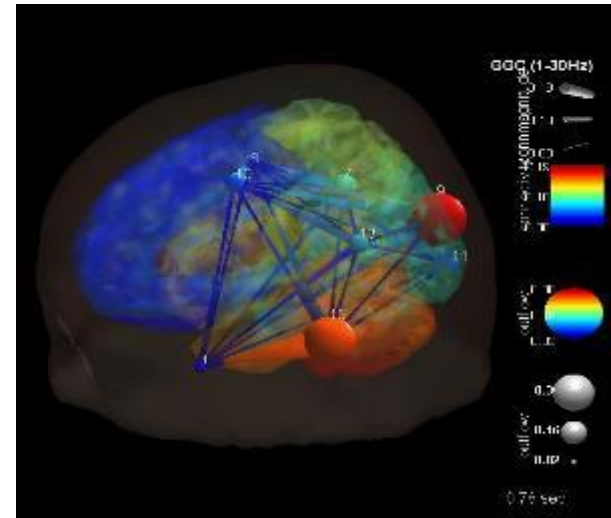
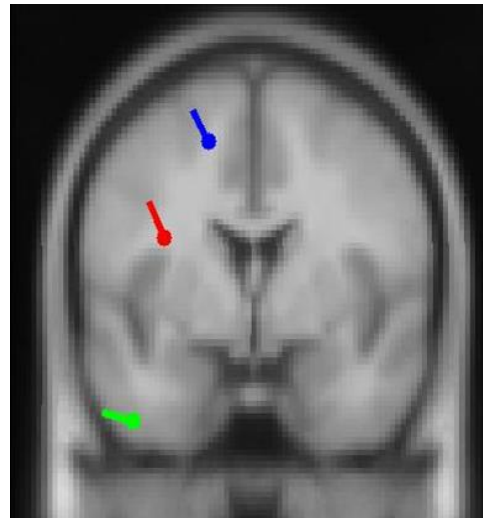


Exemplar: Major Depression

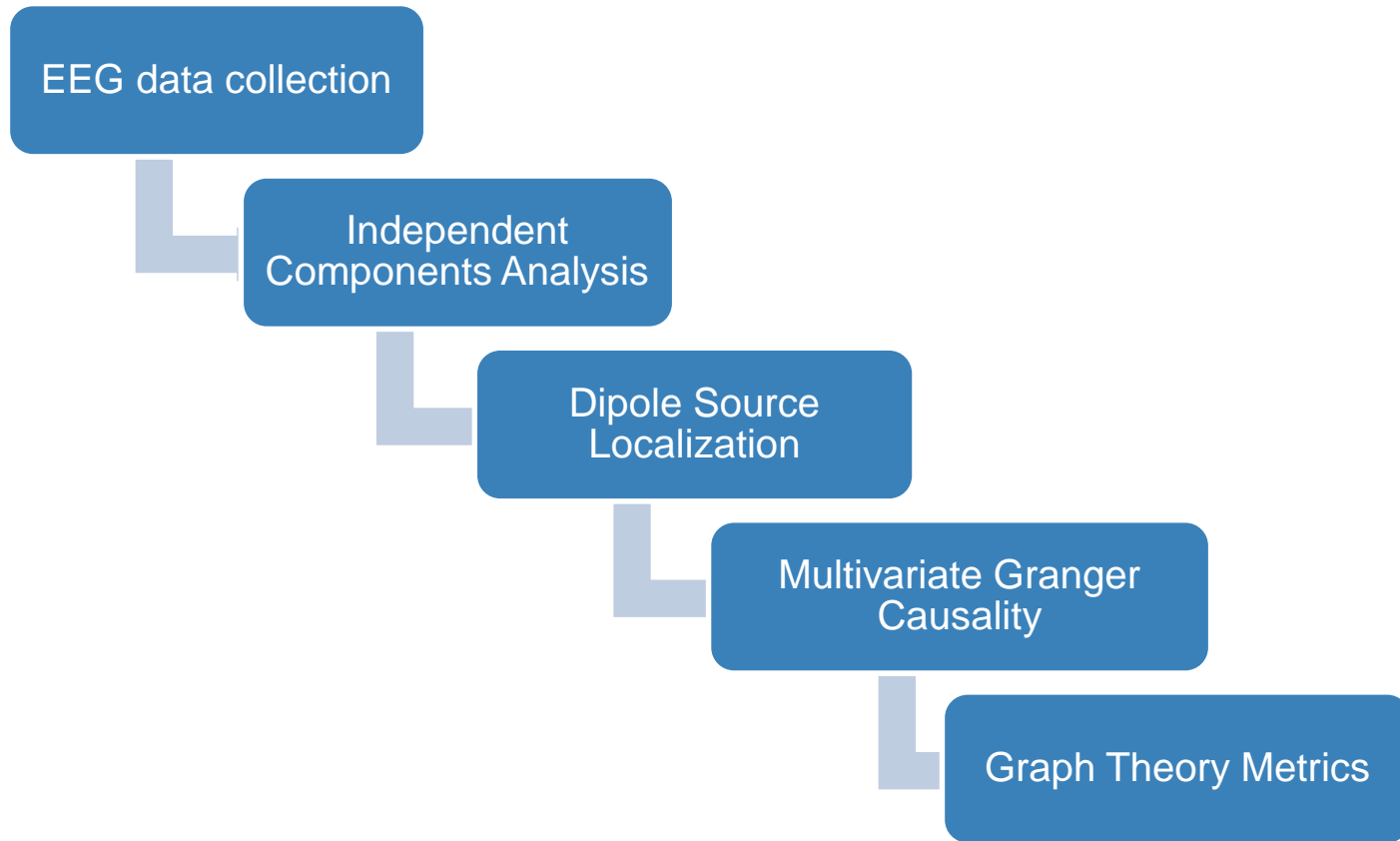
Coherence



EDF file - 1 s epochs resampled



Novel EEG Analysis pipeline focused on effective connectivity assessment



ICA/EEGLAB Scientists and Journals

- Journal of Neuroscience Methods
- Plos One
- Computational Intelligence and Neuroscience
- NeuroImage
- Computational Intelligence and Neuroscience
- Frontiers in Neuroscience
- Frontiers in Neural Circuits
- UCSD Swartz Center for Computational Neuroscience
- University of Oxford
- UCLA Semmel Neuroscience Institute
- MGH/Harvard Medical School
- Georgetown University Medical Center
- University of Michigan Neuroscience Department

EEGLAB: an open source toolbox for analysis of single-trial EEG dynamics including independent component analysis

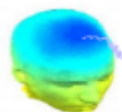
Arnaud Delorme  , Scott Makeig 

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<https://doi.org/10.1016/j.jneumeth.2003.10.009>

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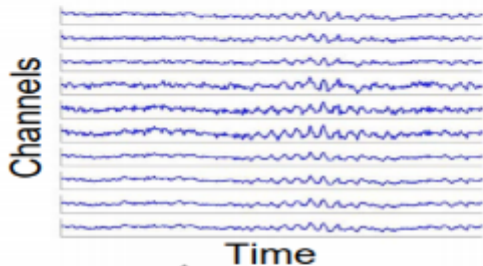
Independent Component Analysis



x = scalp EEG

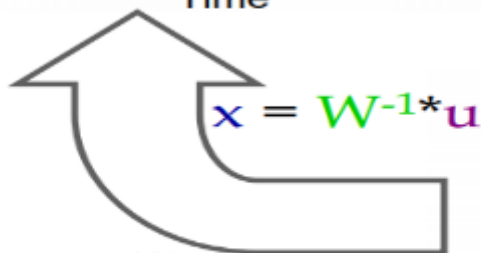
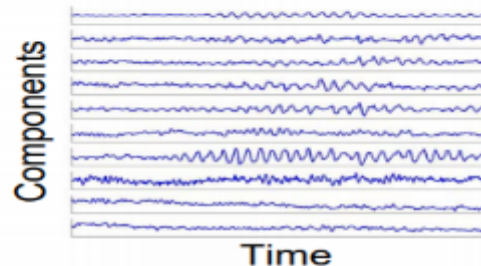
W = unmixing matrix

u = sources



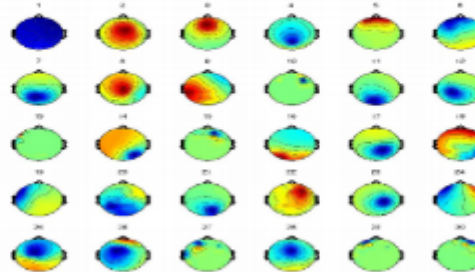
$$W^*x = u$$

ICA

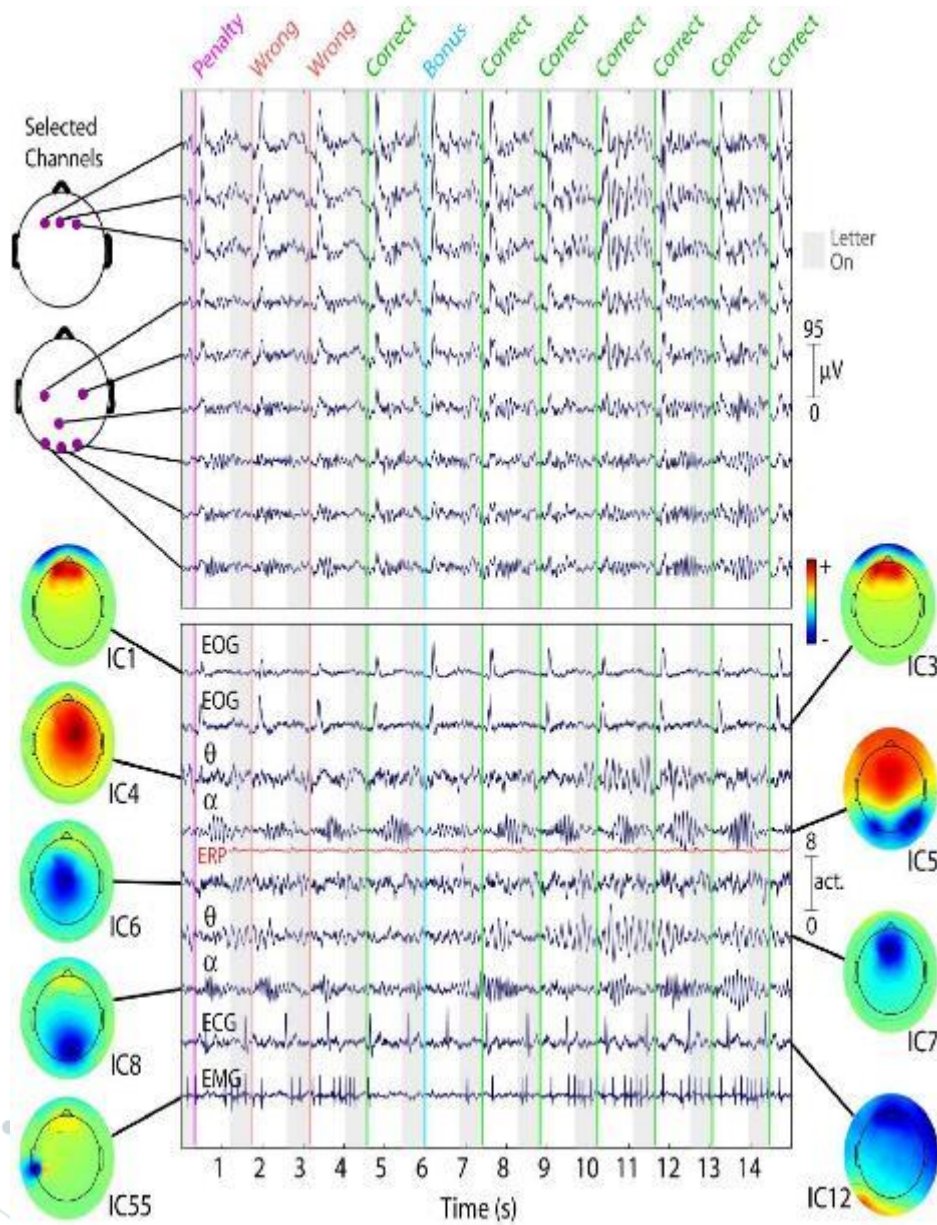


u = sources

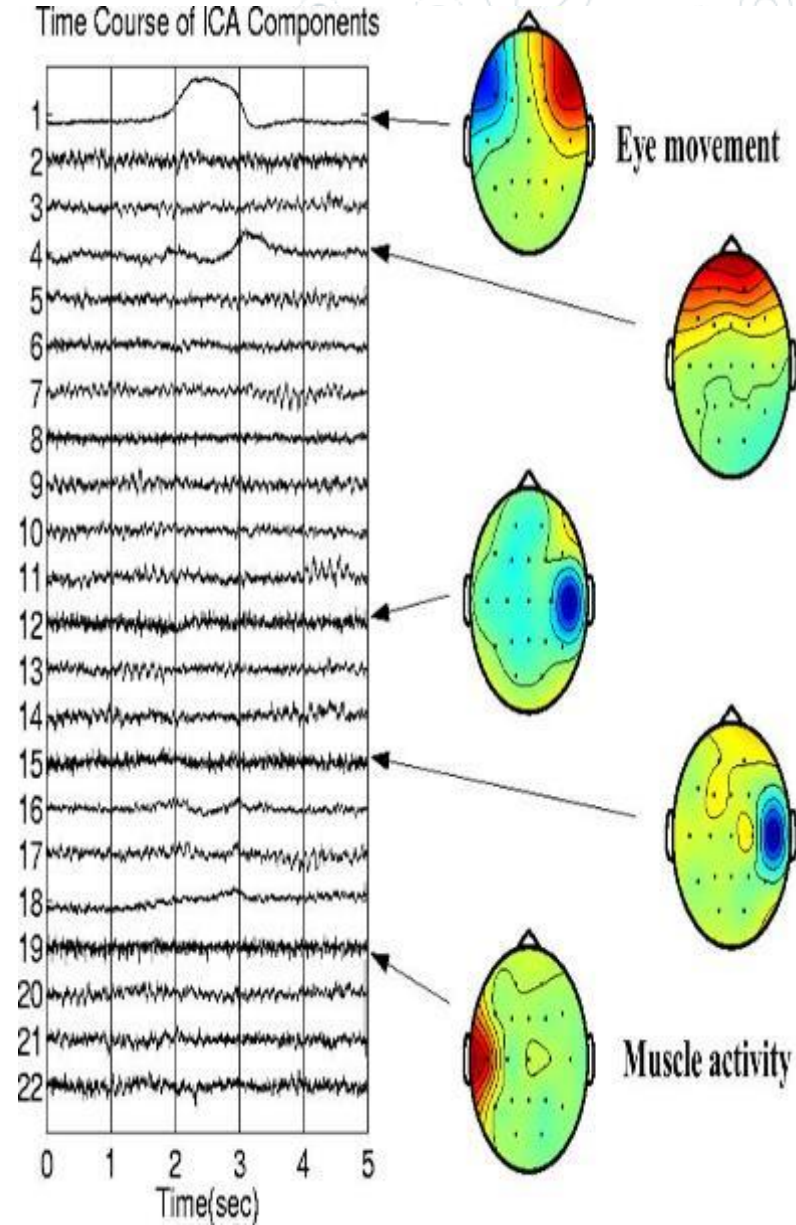
W^{-1} (scalp projections)



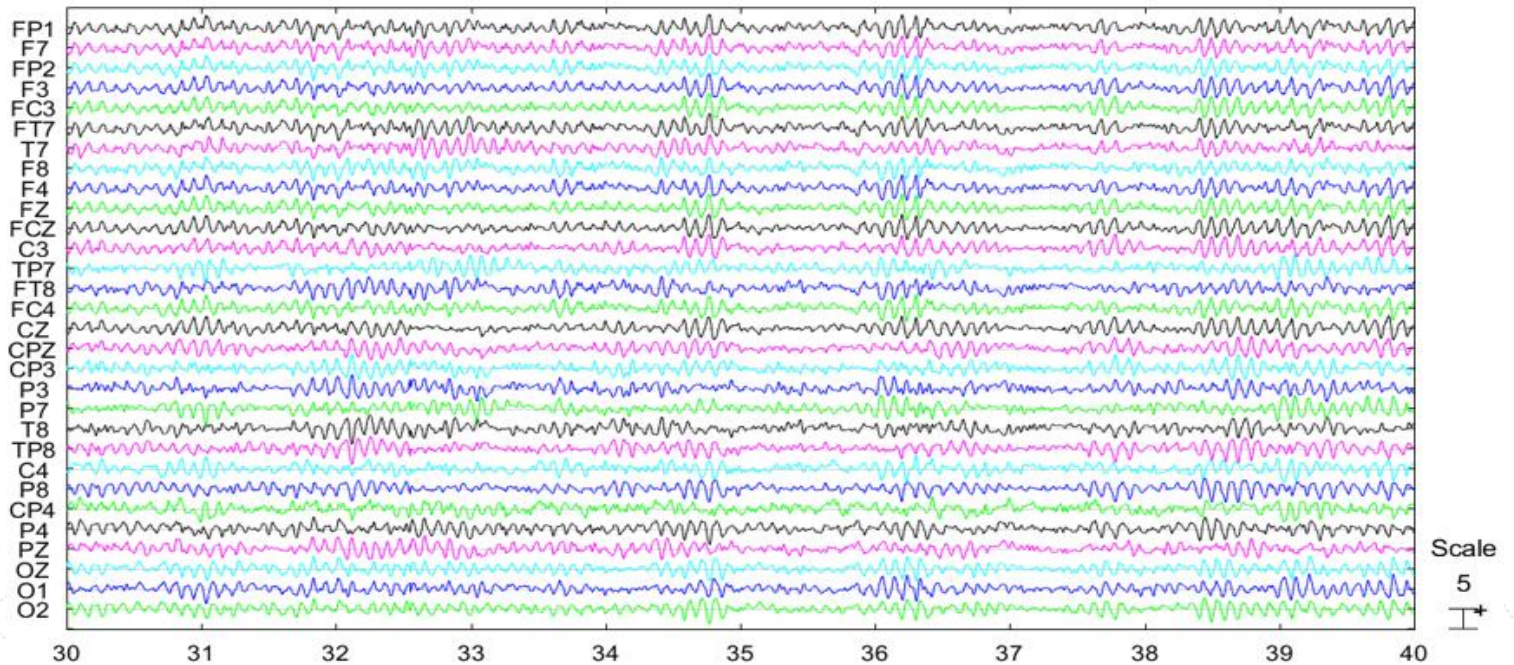
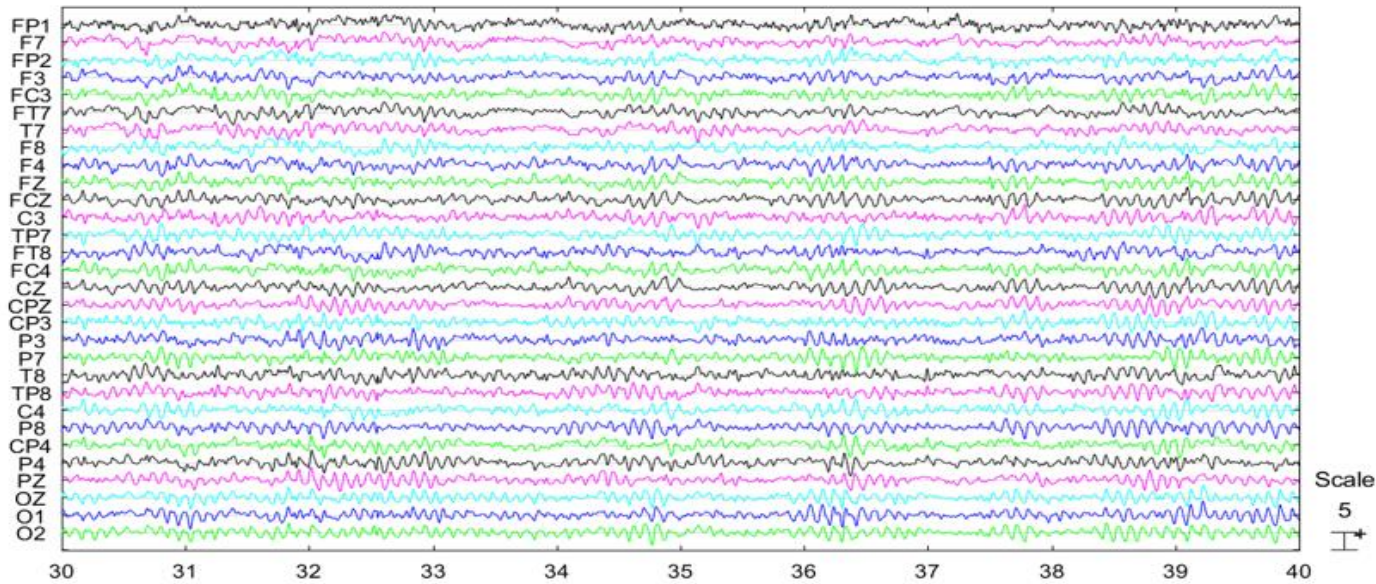
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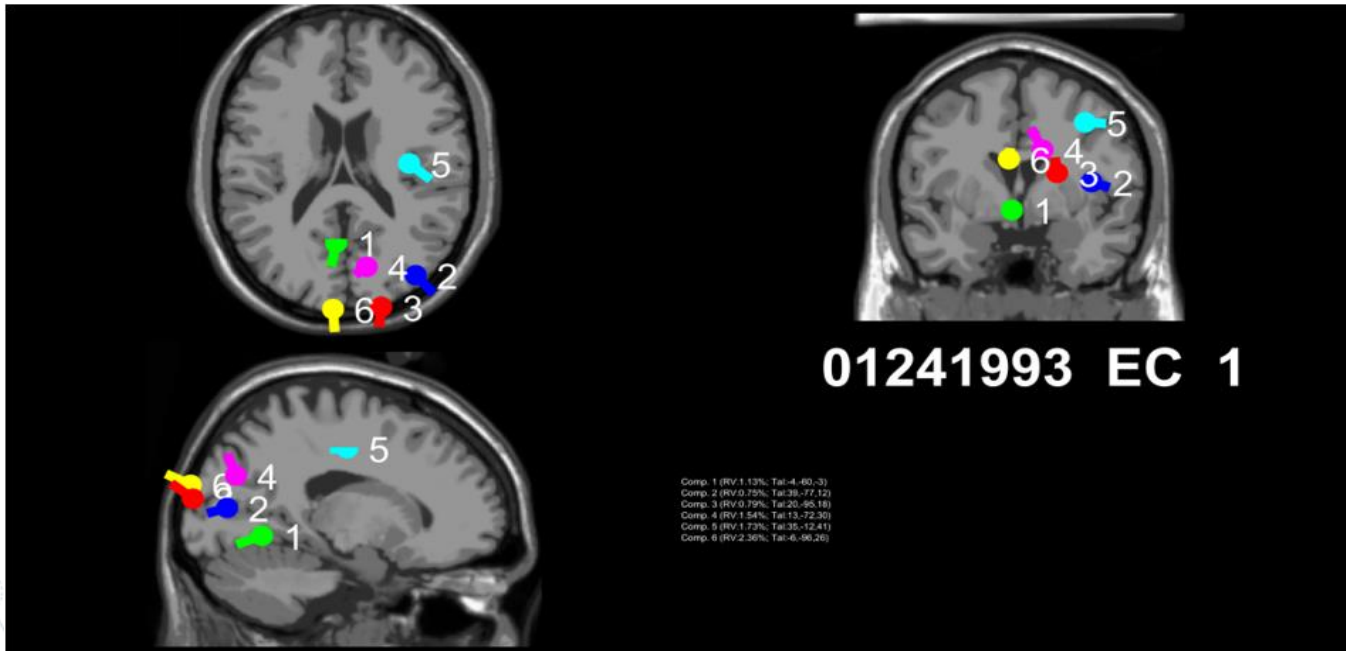
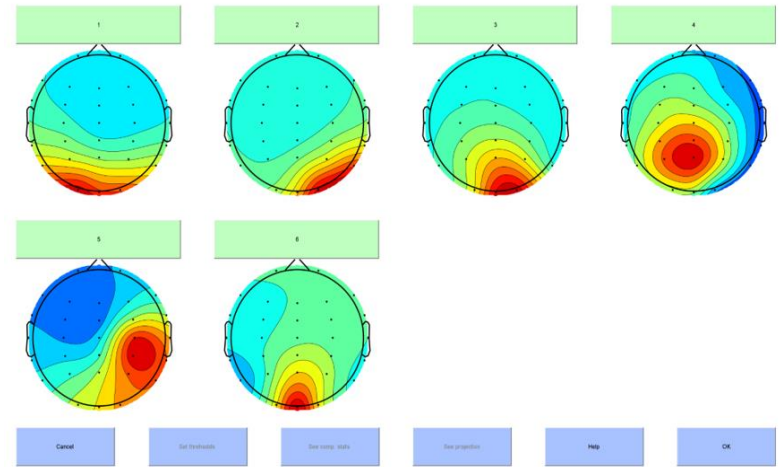
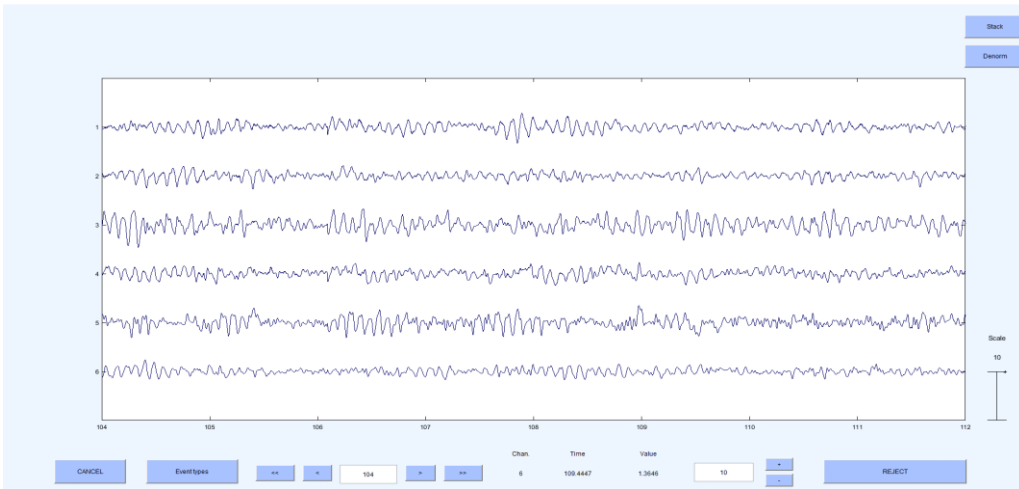


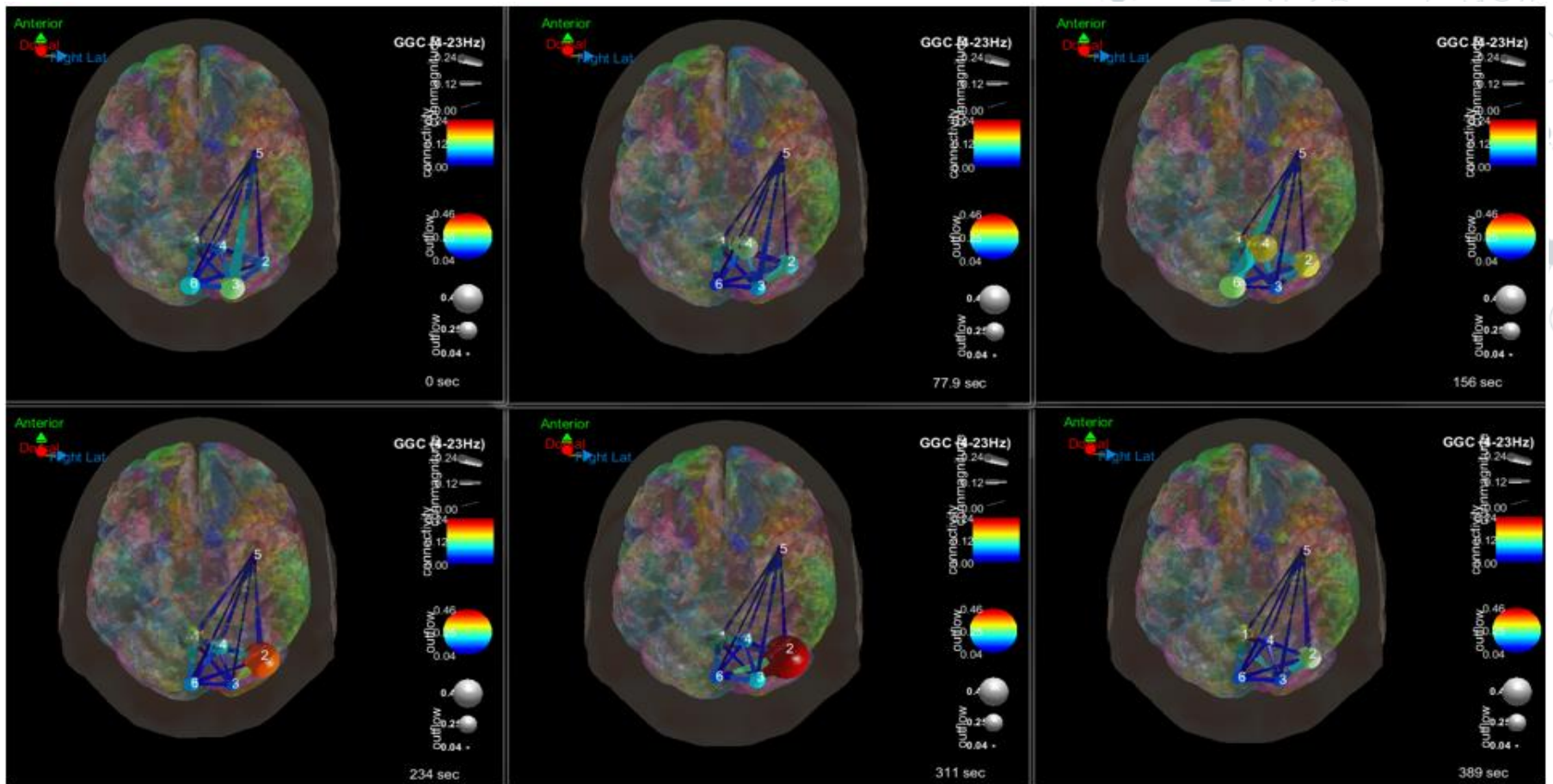
Time Course of ICA Components



Example: 25 year old woman with a history of emotional abuse as a child and adult. Presents with anxiety, panic, nightmares and dissociation.







Graph Theory Metrics

Time Range	Freq Range	Clust Coeff	Path Length	Global Eff	Radius	Diameter
281-352s	4-23Hz	0.031157467	37.27736431	0.044989651	39.41305	91.9196607

Graph Theory: Network Dynamics

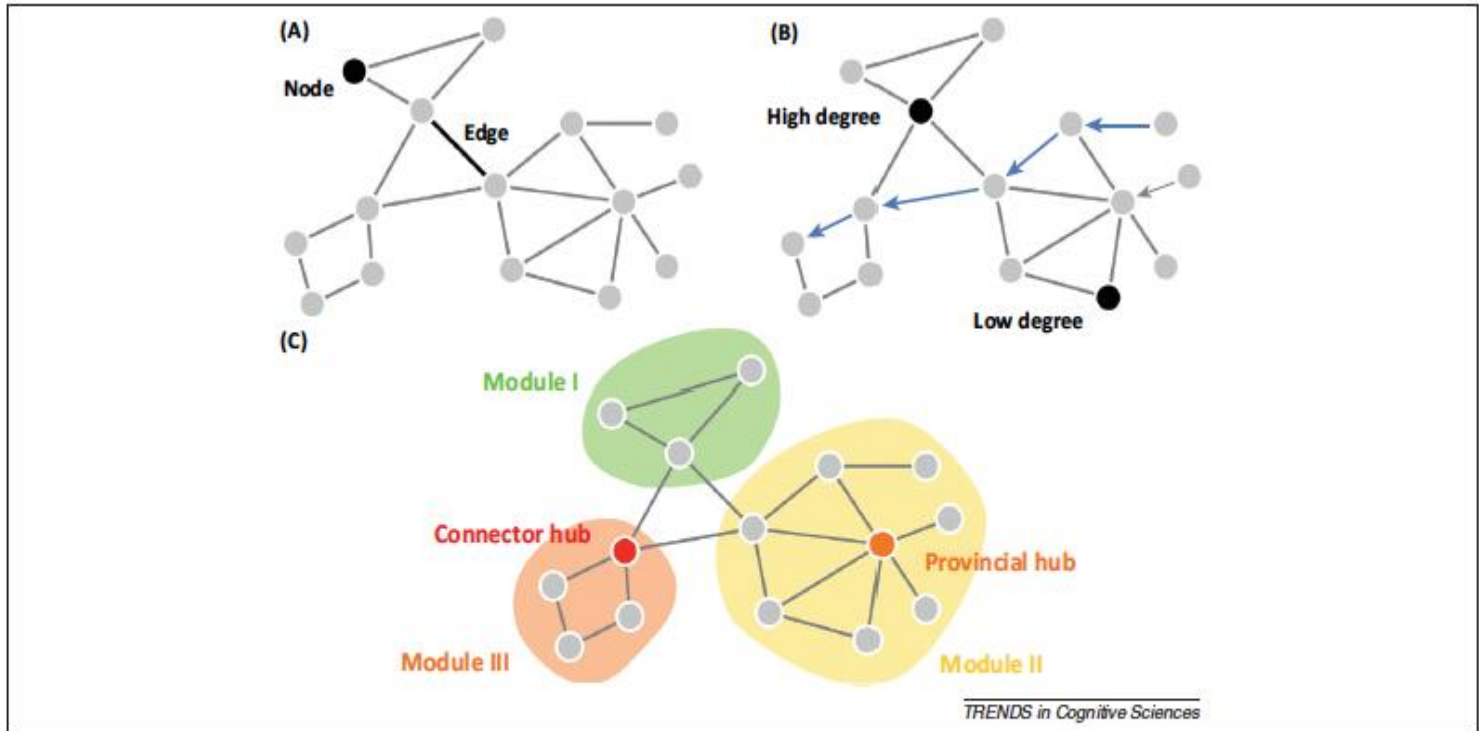
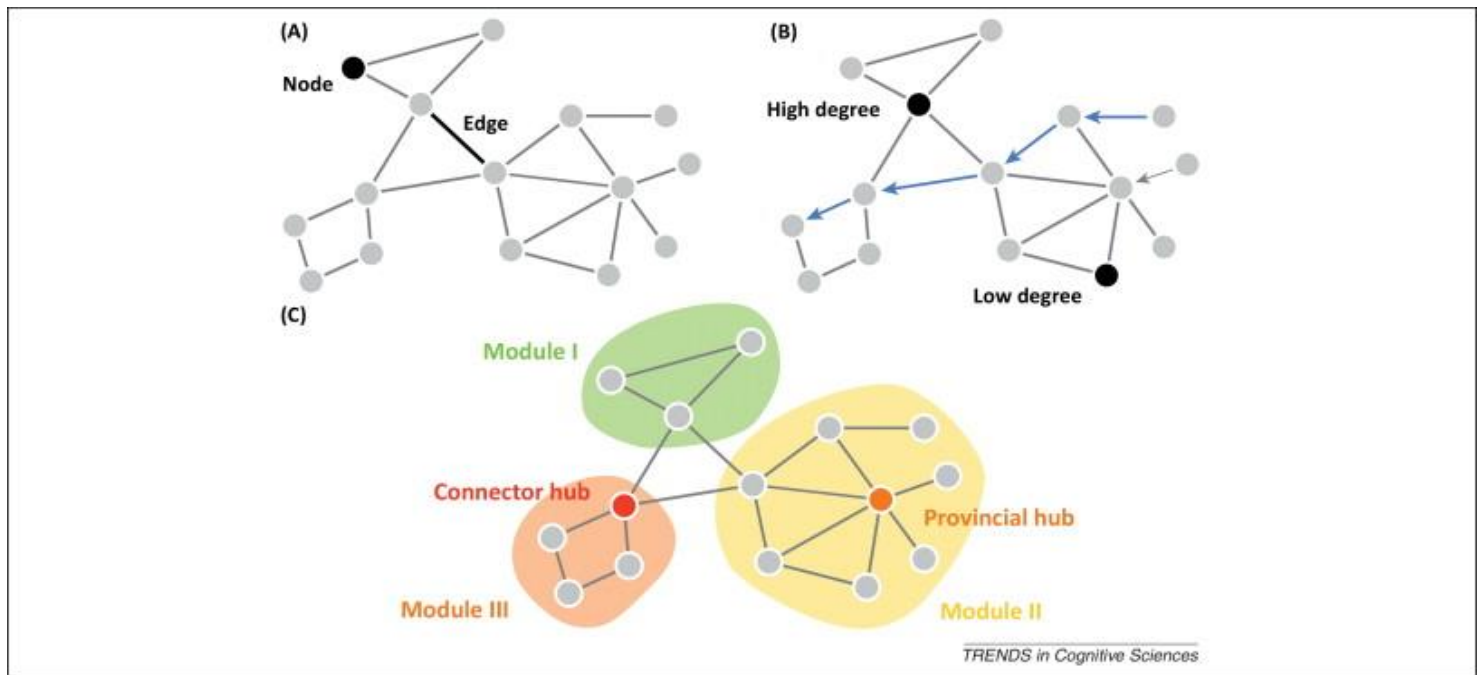


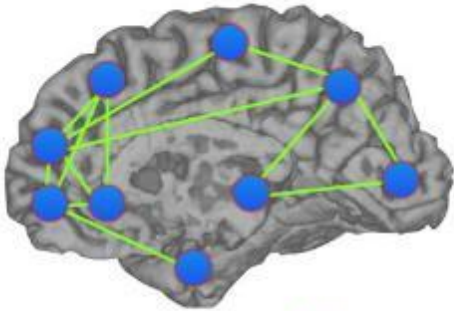
Figure 1. Basic network attributes. **(A)** Brain networks can be described and analyzed as graphs comprising a collection of nodes (describing neurons/brain regions) and a collection of edges (describing structural connections or functional relationships). The arrangement of nodes and edges defines the topological organization of the network. **(B)** A path corresponds to a sequence of unique edges that are crossed when traveling between two nodes in the network. Low-degree nodes are nodes that have a relatively low number of edges; high-degree nodes (often referred to as hubs) are nodes that have a relatively high number of edges. **(C)** A module includes a subset of nodes of the network that show a relatively high level of within-module connectivity and a relatively low level of intermodule connectivity. 'Provincial hubs' are high-degree nodes that primarily connect to nodes in the same module. 'Connector hubs' are high-degree nodes that show a diverse connectivity profile by connecting to several different modules within the network.

Graph Elements

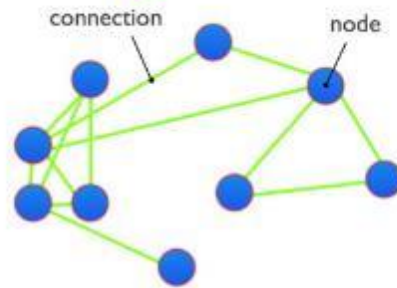
- ⦿ Hubs: A node with links that exceeds average.
- ⦿ Low vs High Degree Hubs.
- ⦿ Provincial Hubs.
- ⦿ Modules.
- ⦿ Connector Hubs.



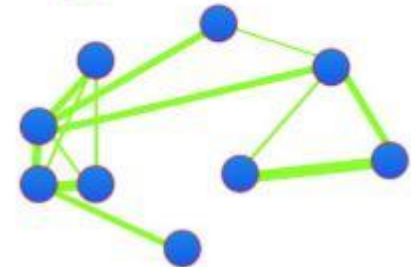
a the structural brain network



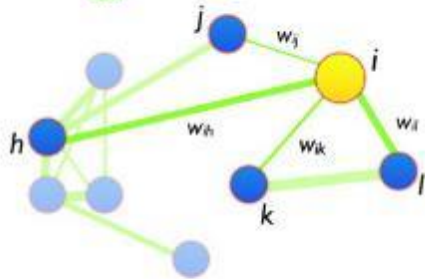
b graph



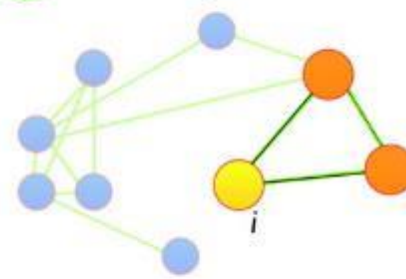
c weighted graph



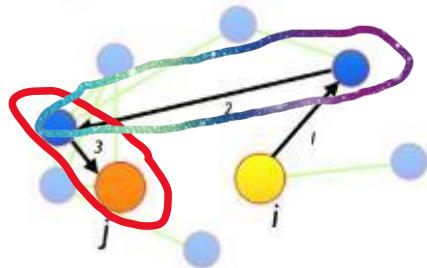
d strength S_i



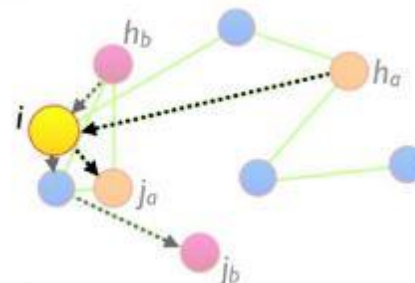
e clustering-coefficient C_i



f path length L_i



g betweenness centrality B_i



Radius **Diameter**

REVIEW ARTICLE

The functional brain connectome of the child and autism spectrum disorders

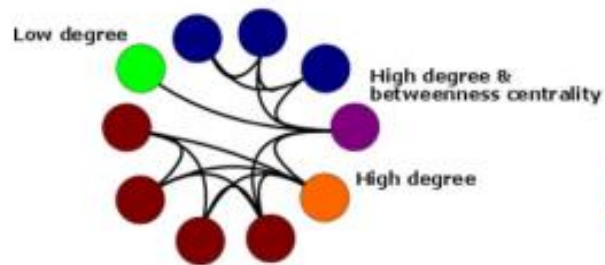
Katell Mevel¹, Peter Fransson (Peter.Fransson@ki.se)²

1.Laboratory for the Psychology of Child Development and Education (LaPsyDÉ), CNRS UMR 8240, Sorbonne Paris Cité, GIP Cyceron, Université de Caen Normandie, Université Paris Descartes, Paris, France

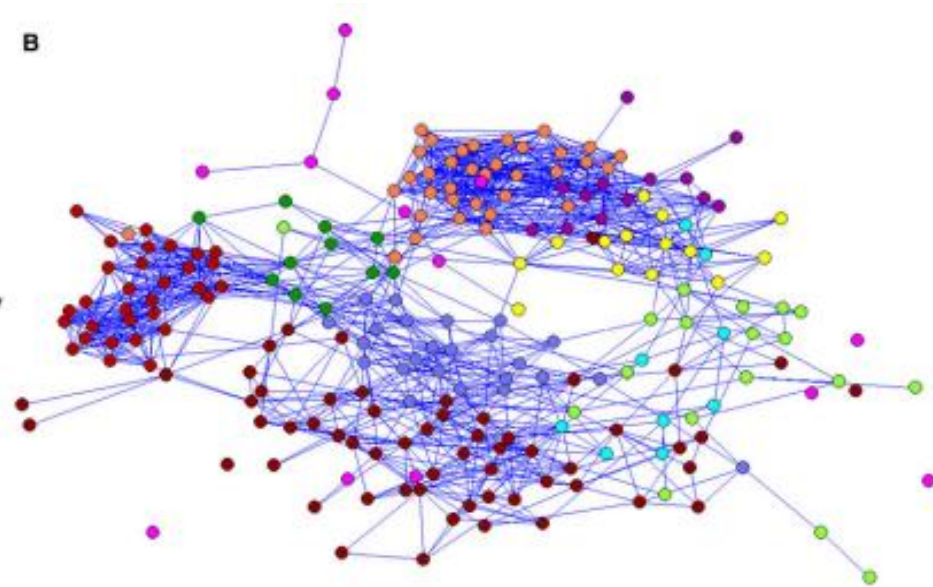
2.Department of Clinical Neuroscience, Karolinska Institutet, Stockholm, Sweden

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A



B



DM	Vis	Sa	Au	VA
SM	FP	CO	Sub	DA

Four Channel Multivariate Coherence Training: Development and Evidence in Support of a New Form of Neurofeedback

Robert Coben^{1*}, Morgan Middlebrooks¹, Howard Lightstone² and Madeleine Corbell³

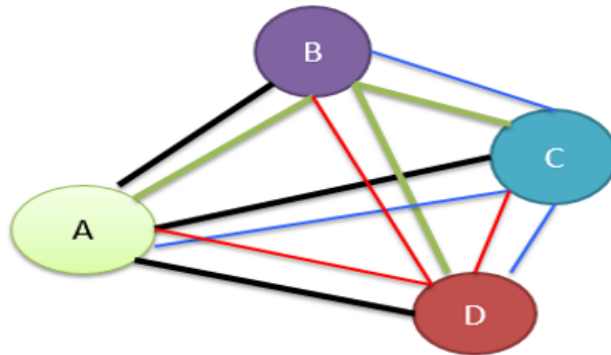
¹Integrated Neuroscience Services, Fayetteville, AR, United States

²EEG Software, LLC, Gainesville, FL, United States

³Department of Psychological Science, University of Arkansas, Fayetteville, AR, United States

QPS: Averaging coherences

- A method of combining averaged psync values.
 - 4 channels of EEG
 - Each pair has a running psync calculation
 - For each channel, the 3 pairs of psync values are computed, averaged and this is used as the output reward value
 - If a raw channel is in artifact condition, the channel is not used in the averaging calculation



$$A = (AB + AC + AD)/3$$

$$B = (BA + BC + BD)/3$$

$$C = (CA + CB + CD)/3$$

$$D = (DA + DB + DC)/3$$

$$\text{QPS Ave} = (A + B + C + D)/4$$

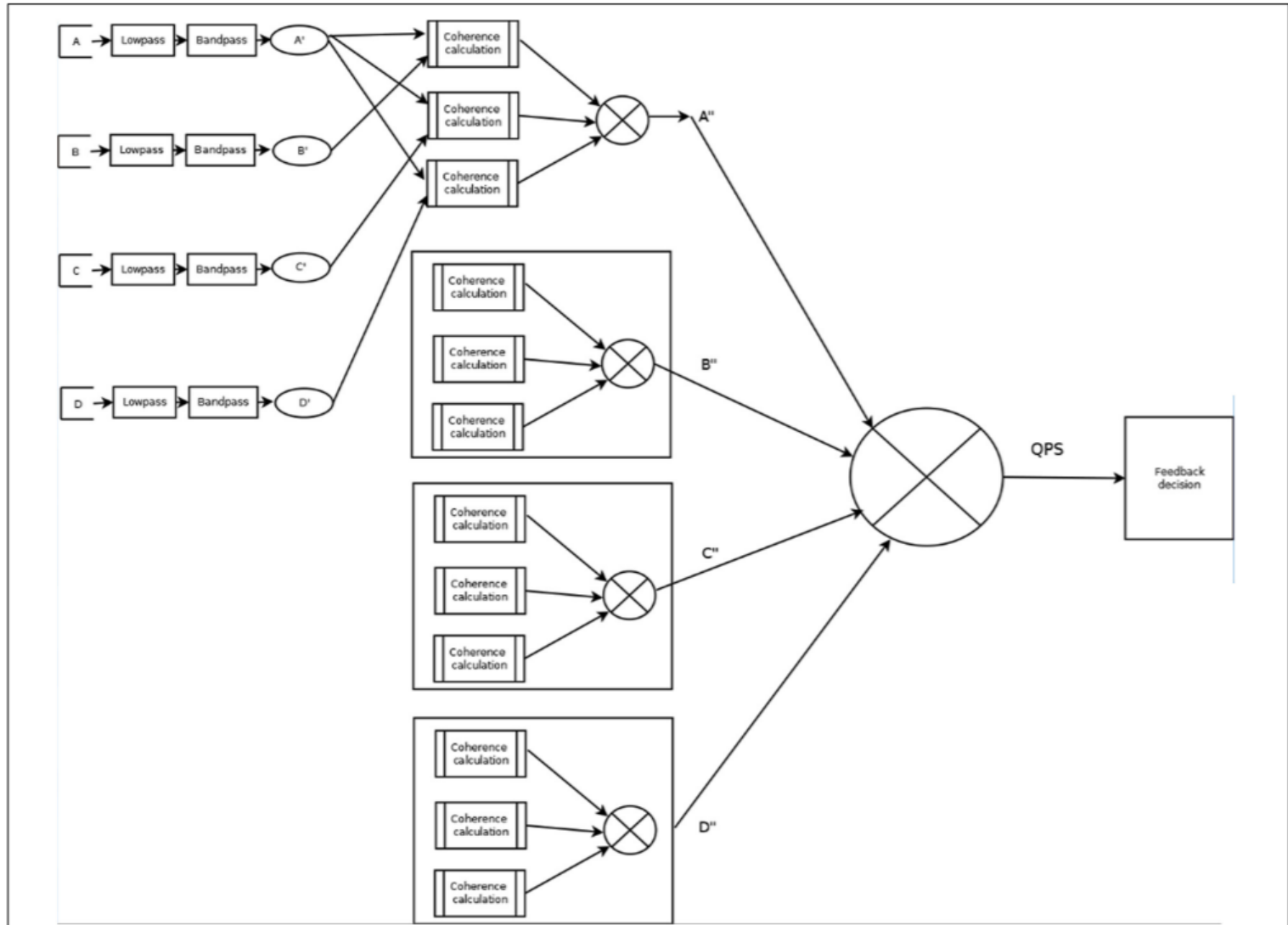


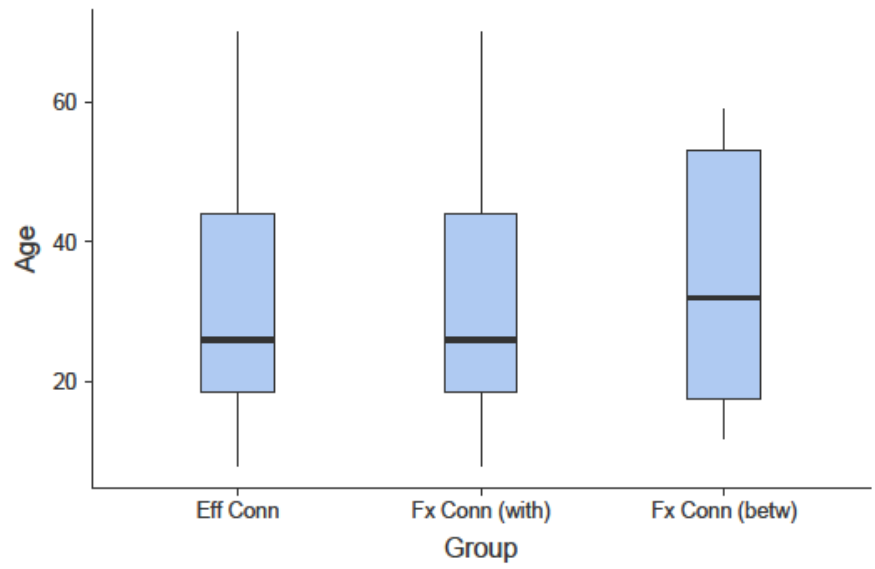
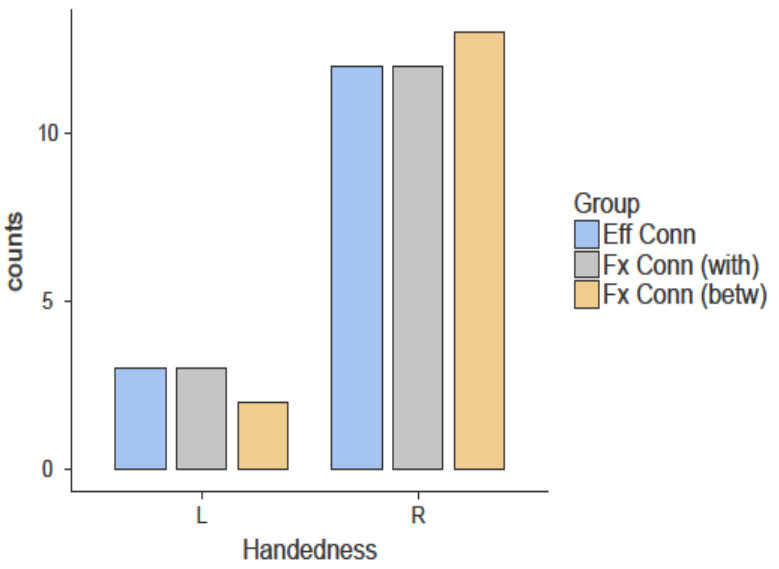
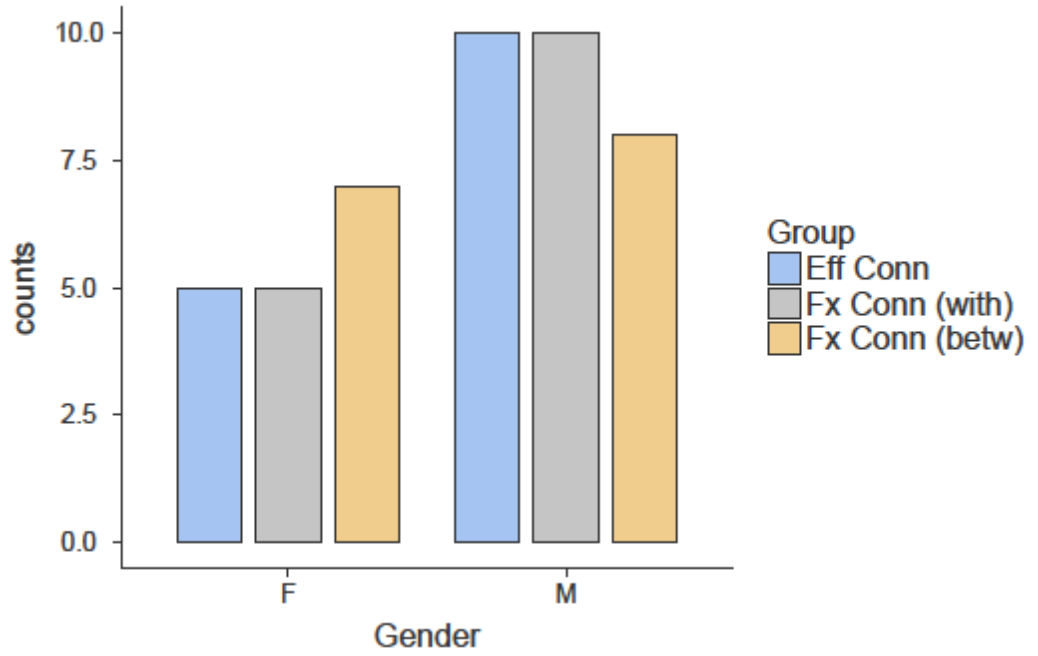
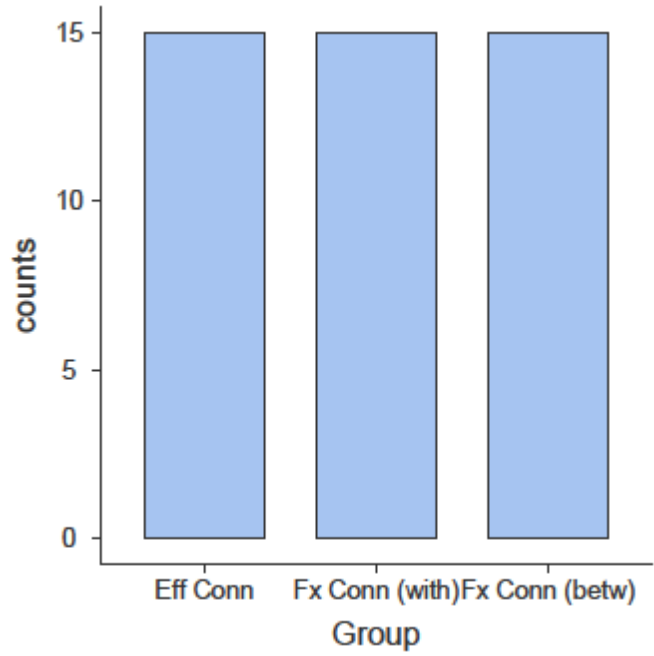
FIGURE 2 | Visual representation of A_Psync filter.

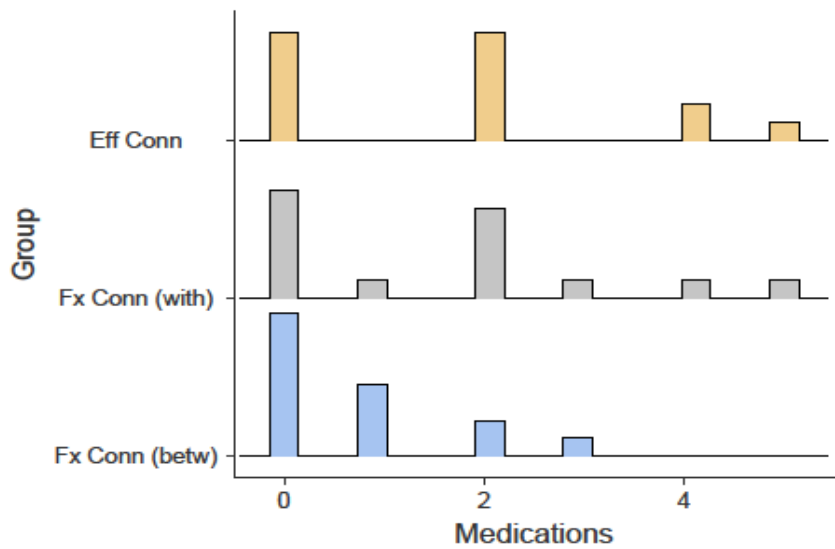
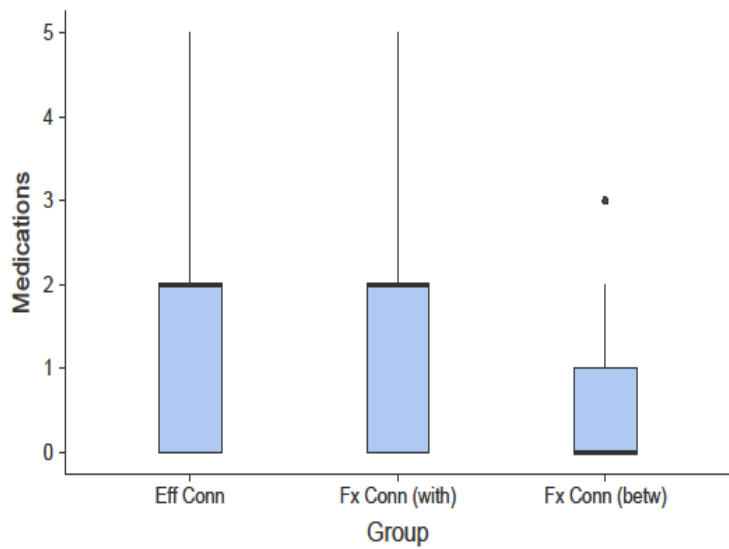
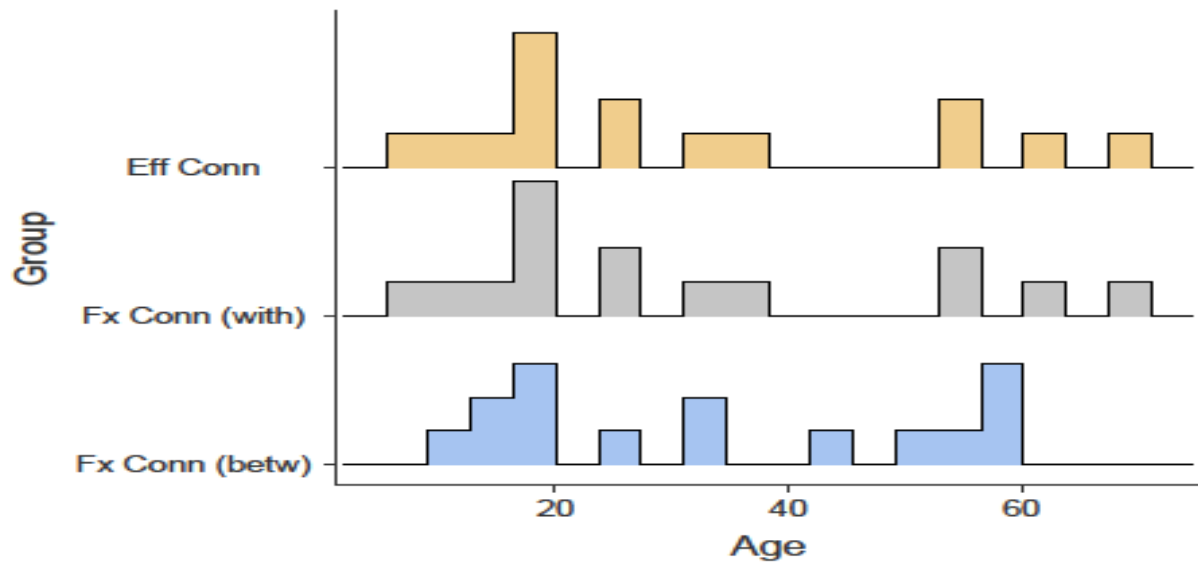
Efficacy Studies in Support of 4 channel MVCNF (N = 591)

Population	Sample	Design	Findings 1	Findings 2
General Population	N = 174	MVCNF v 2 Ch CNF	MVCNF > 2 Ch CNF	Enhanced coherence and reduced power
Traumatic Brain Injury	N = 20	Compared time since injury in 3 groups	Improvements in symptoms and NP testing	Changes associated with increases in coherence
Epilepsy	N = 52	MVCNF v 2 Ch CNF	MVCNF > 2 Ch CNF	81% reduction in seizures
Learning Disabilities	N = 63	MVCNF v 2 ch CNF v resource room	MVCNF > 2 ChCNF > RR	1.6 year increase in reading
Autism	N = 110	MVCNF v 2 Ch CNF	MVCNF > 2 Ch CNF	98% success rate
Autism MND	N = 78	MVCNF v 2 Ch CNF v Bipolar	MVCNF > 2 ChCNF > Bipolar	Mu suppression with coherence changes
Depression	N = 54	MVCNF Psychotherapy v WLC	MVCNF > both groups	94% success rate, crossover and 2 yr f/u
Developmental Trauma	N = 40	MVCNF v. Psychotherapy	Exp > controls on clinical ratings	Δ in power, sources and connectivity

Study Methodology

- ◎ Subjects were assigned to one of three groups (N = 45).
- ◎ These included an effective connectivity (15), functional connectivity within group (15) and a functional connectivity between group (15) comparison. Group 1 and 2 were the same subjects (within groups) that received different interventions at different time points (FC always first).
- ◎ All subjects received four channel multivariate coherence training over 12-15 sessions.
- ◎ Clinical ratings and therapist ratings (0-20) were derived at the completion of their treatment regimen.
- ◎ Client ratings were largely subjective and based on self-ratings only or parental ratings at the completion of training and during the process.
- ◎ Therapist ratings were performed at the completion of training and were based on objective test findings including neuropsychological, behavioral and qeeg findings that reflected change over time.
- ◎ QEEG analysis of change included measures of power at the component level, dipole sources, spectral properties, and multiple measures of graph theory connectivity.





Statistical comparisons for demographics across groups

Age	Gender	Handedness	Medications
F = 0.123	$X^2 = 0.756$	$X^2 = 0.304$	F = 1.82
$p = 0.884$	$p = 0.685$	$p = 0.859$	$p = 0.174$

ANOVA

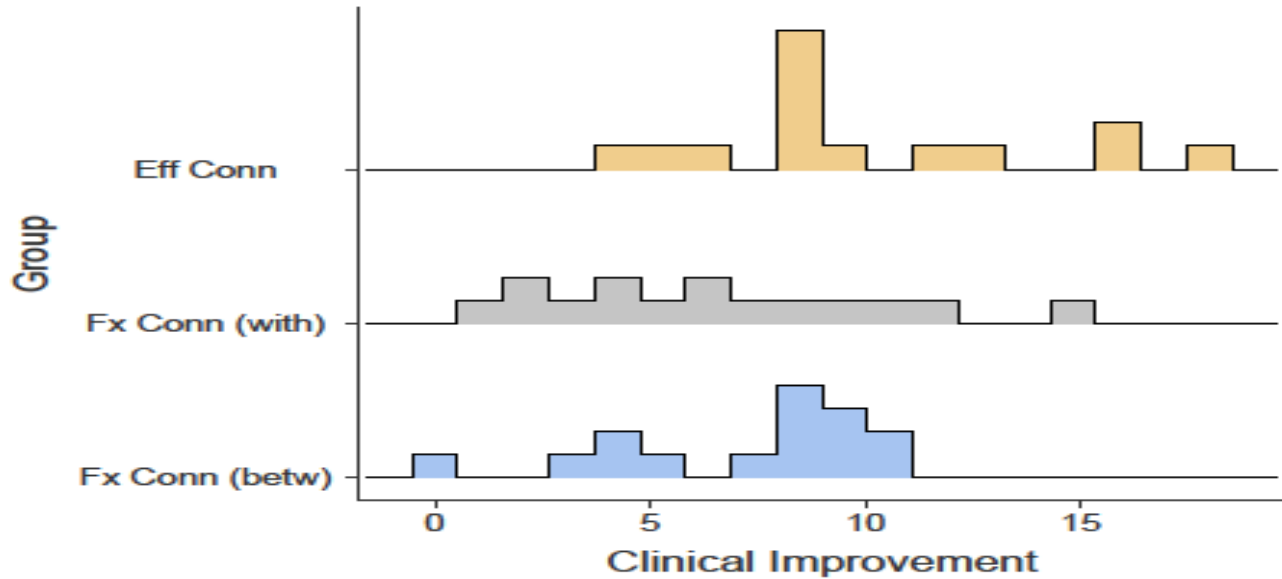
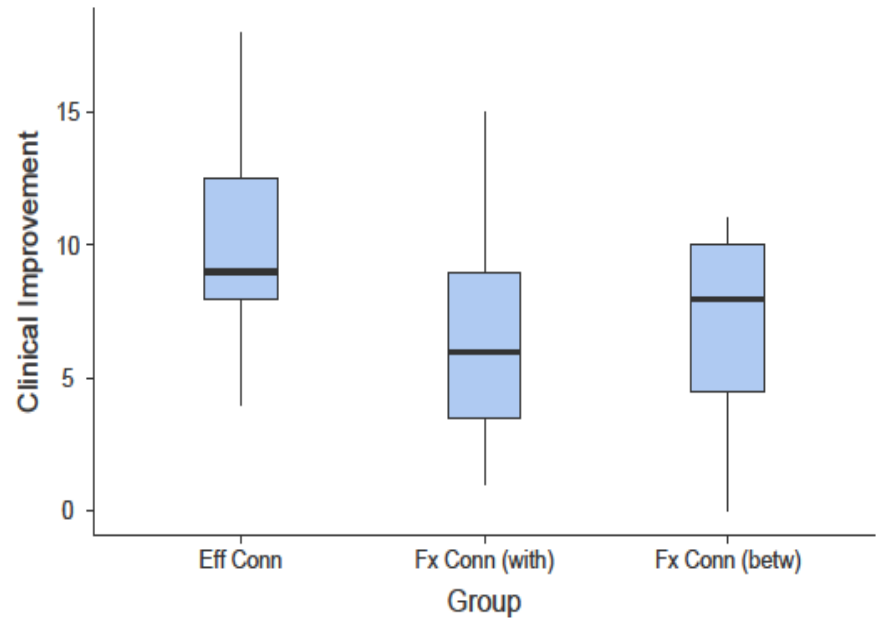
ANOVA

	Sum of Squares	df	Mean Square	F	p
Group	110	2	55.1	3.66	0.034
Residuals	631	42	15.0		

Contrasts

Contrasts - Group

	Estimate	SE	t	p
Fx Conn (with) - Eff Conn	-3.67	1.42	-2.59	0.013
Fx Conn (betw) - Eff Conn	-2.80	1.42	-1.98	0.055



ANOVA

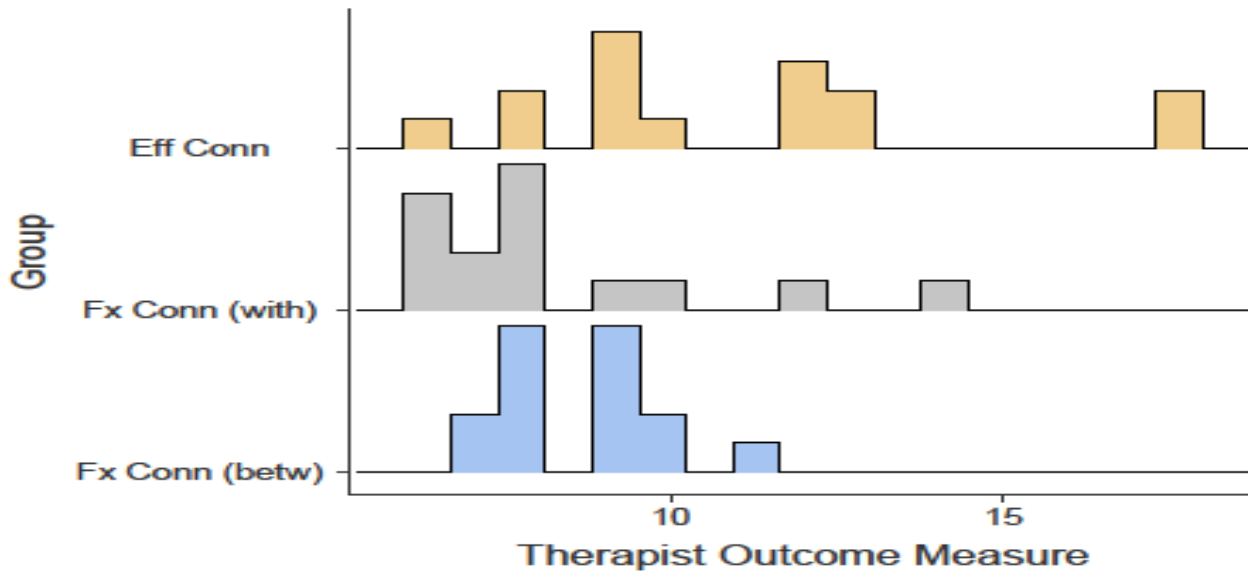
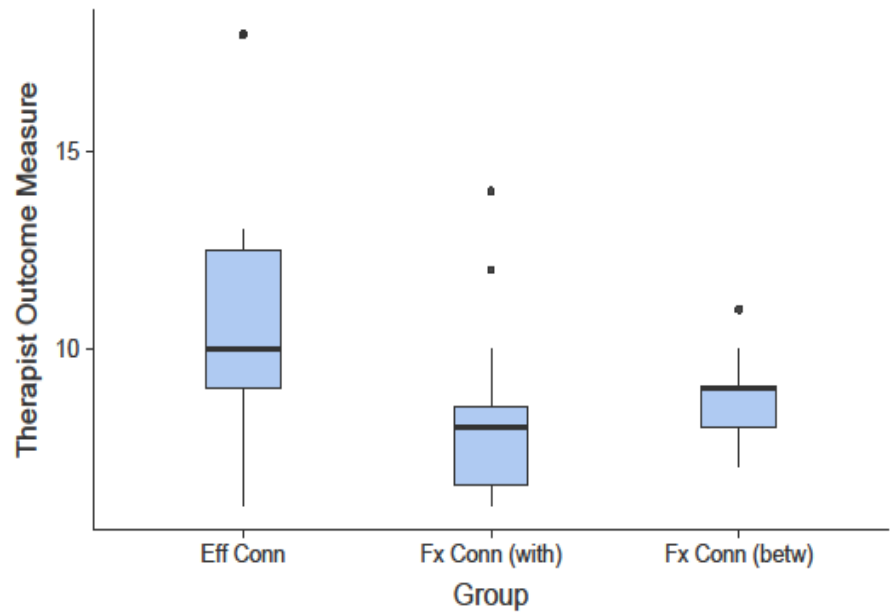
ANOVA

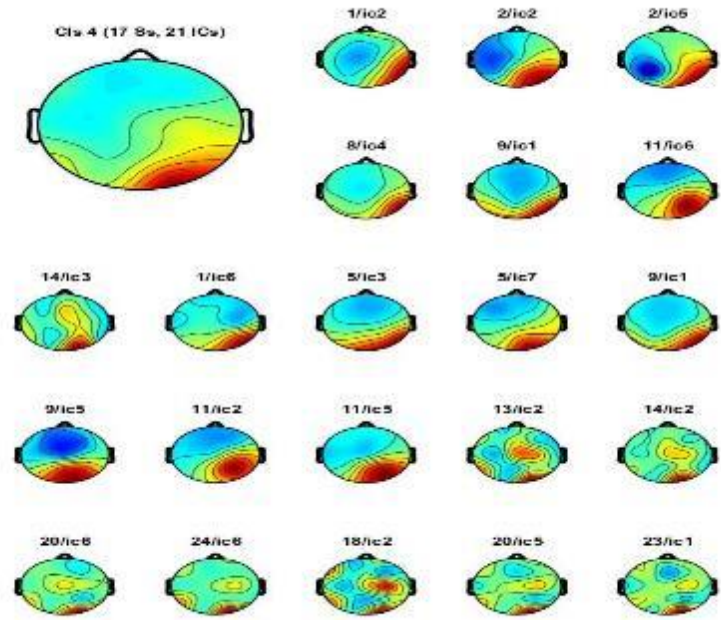
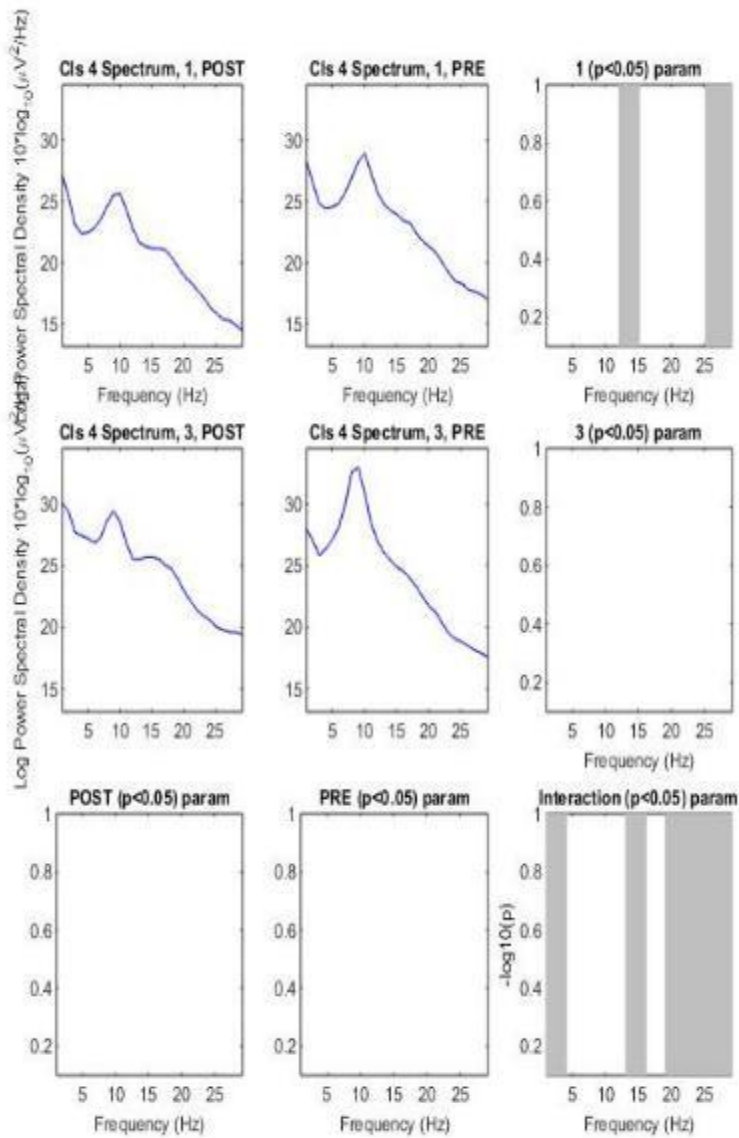
	Sum of Squares	df	Mean Square	F	p
Group	71.0	2	35.49	5.72	0.006
Residuals	260.7	42	6.21		

Contrasts

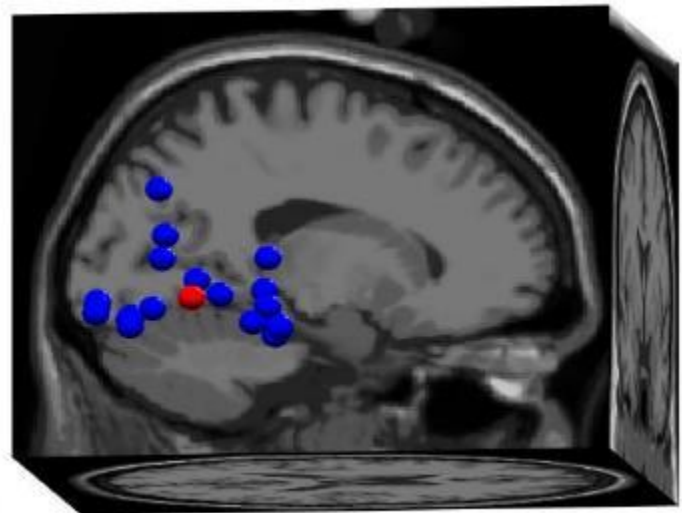
Contrasts - Group

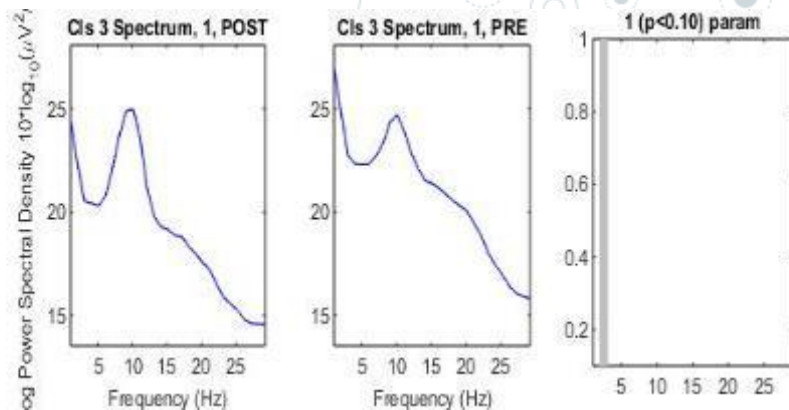
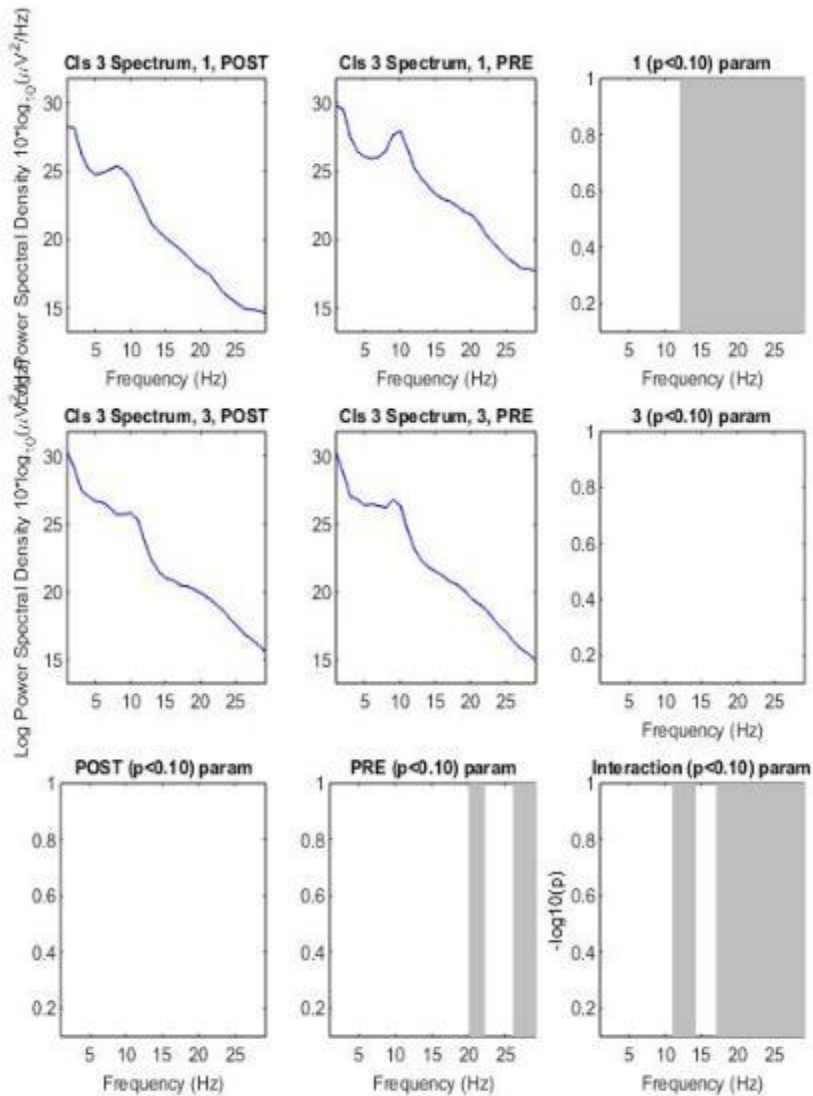
	Estimate	SE	t	p
Fx Conn (with) - Eff Conn	-2.87	0.910	-3.15	0.003
Fx Conn (betw) - Eff Conn	-2.40	0.910	-2.64	0.012





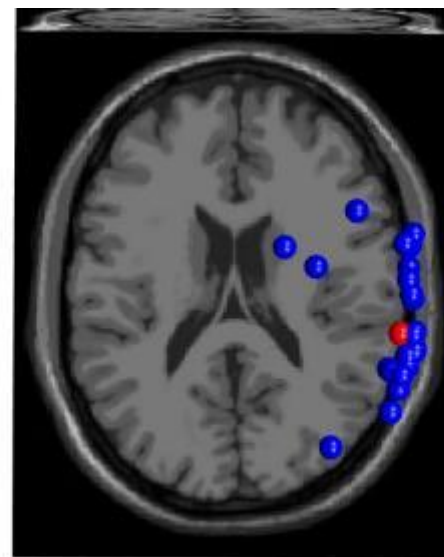
- 22 dipoles:**
- Plot one
 - Keep|Next
 - Next
 - Prev
 - Keep|Prev
 - 1
 - 1, IC2
 - RV: 6.55%
 - X tal: 70
 - Y tal: -49
 - Z tal: 4
 - Display:**
 - Mesh on
 - Tight view
 - Sagittal view
 - Coronal view
 - Top view
 - No controls**

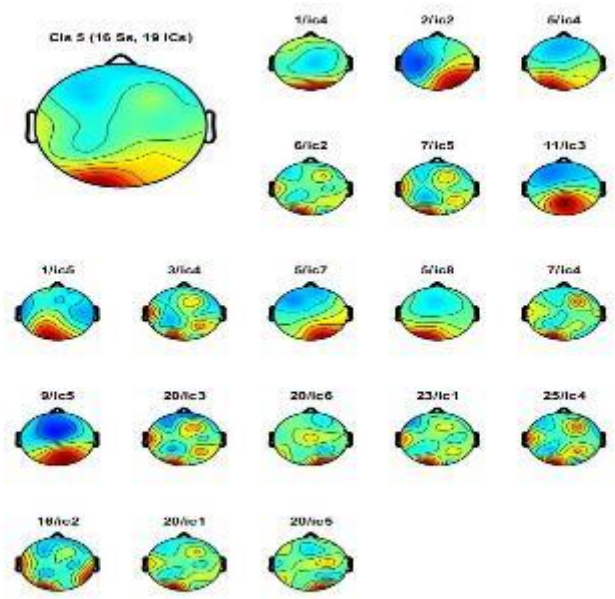
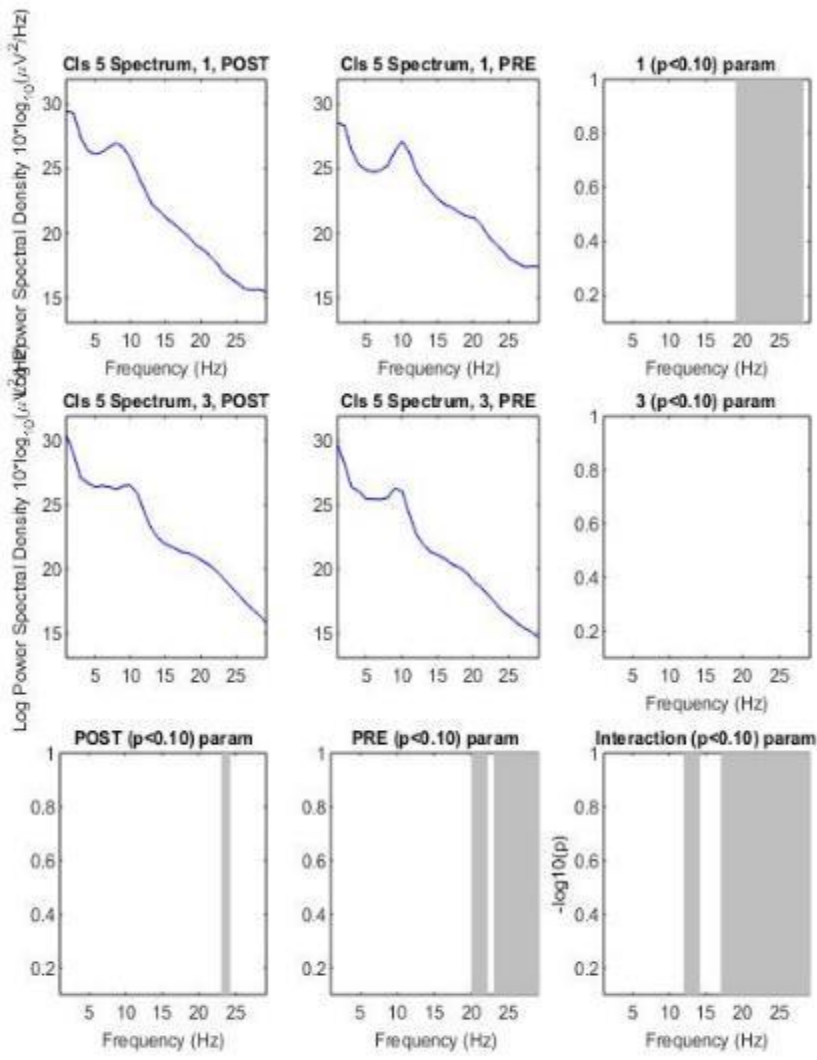




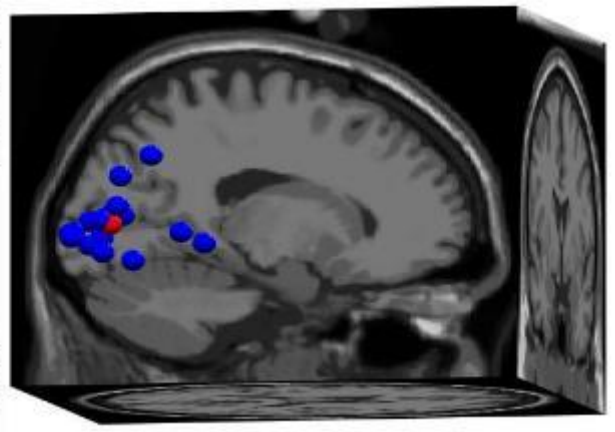
41 dipoles:

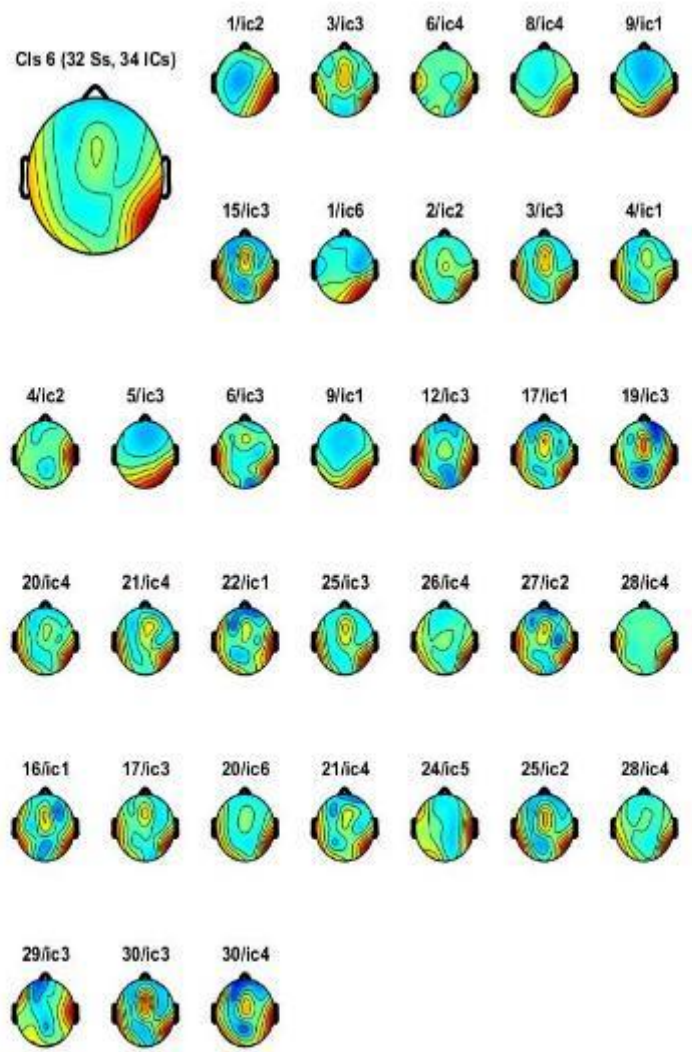
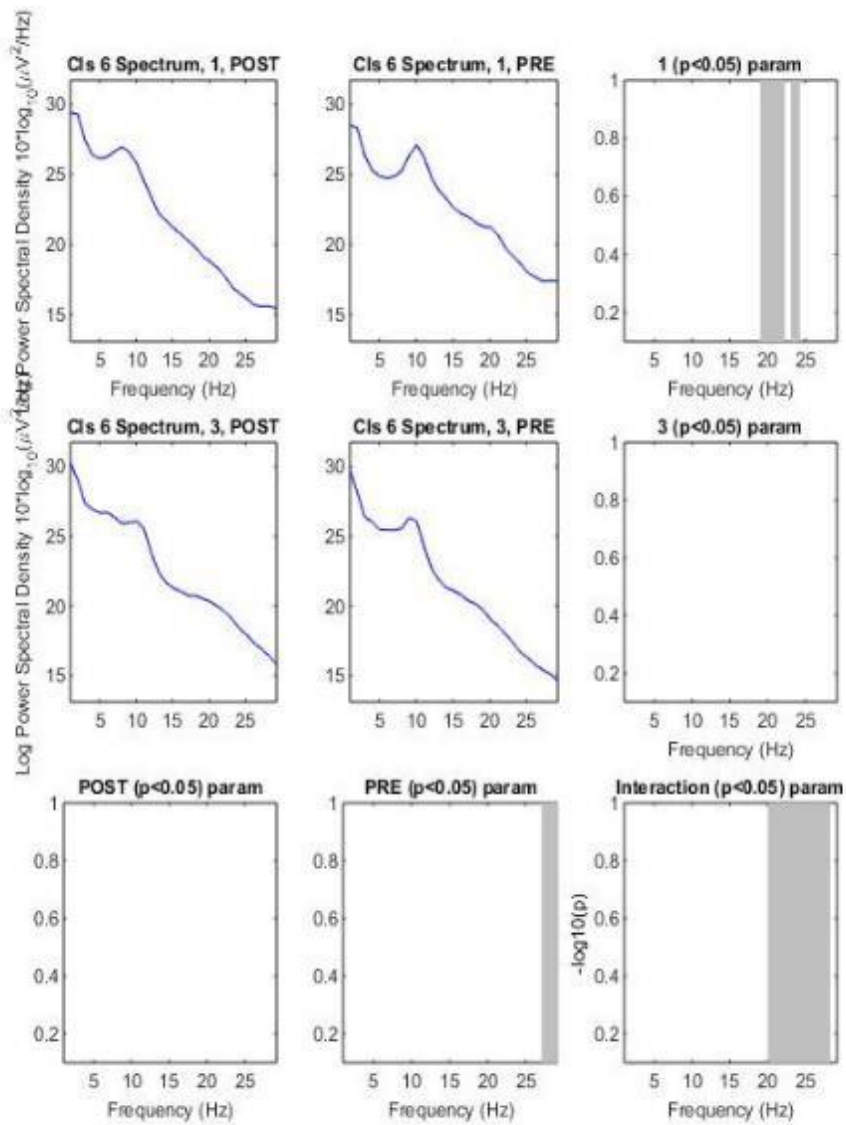
- Plot one
- Keep/Next
- Next
- Prev
- Keep/Prev
- 1
- 1, IC2
- RV: 6.55%
- X tal: 70
- Y tal: -49
- Z tal: 4
- Display:**
- Mesh on
- Tight view
- Sagittal view
- Coronal view
- Top view
- No controls**

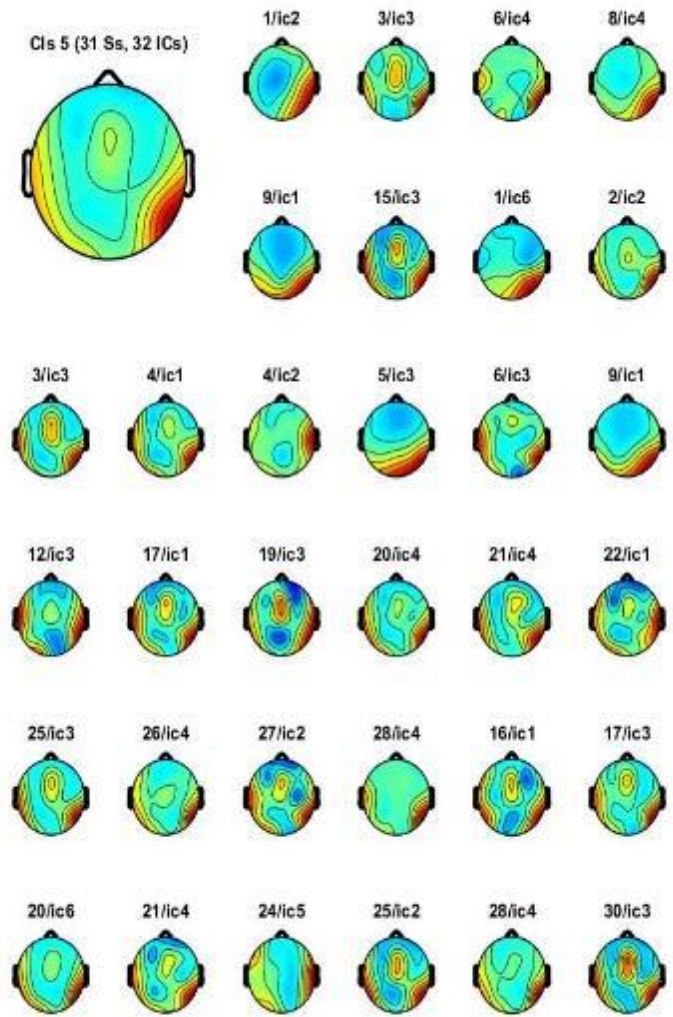
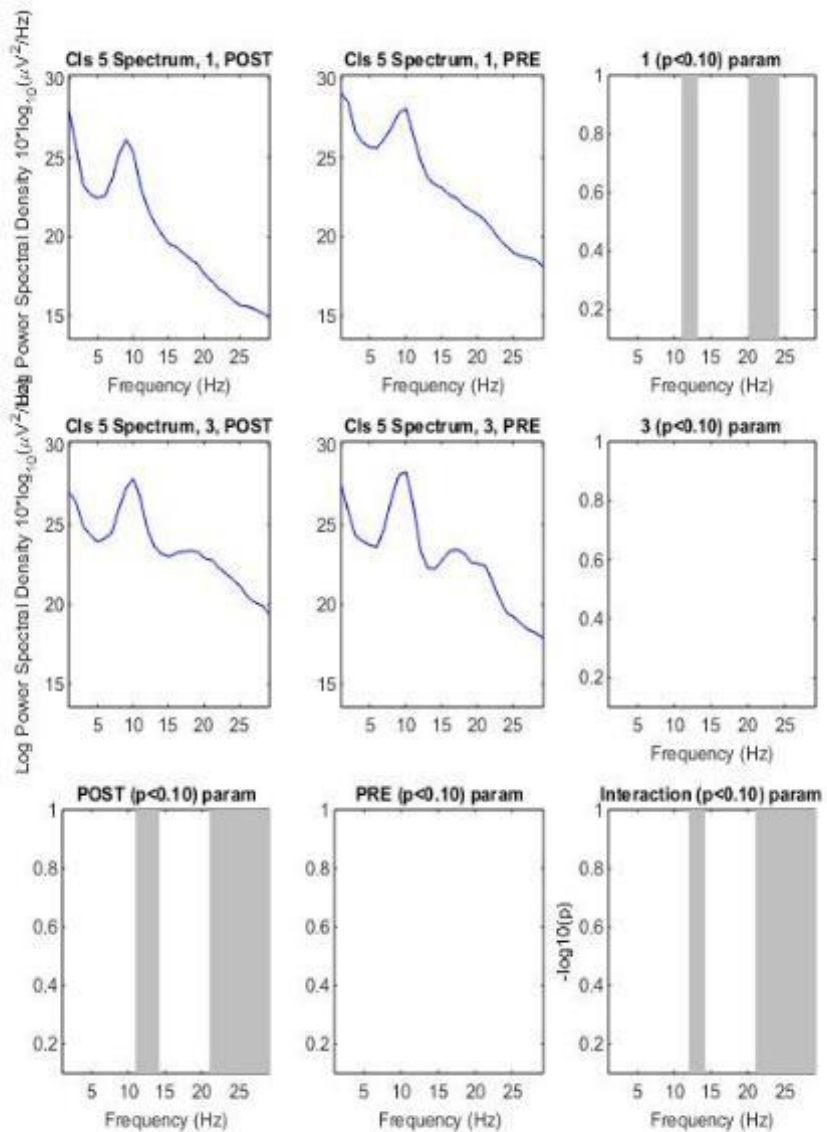




- 20 dipoles:**
- Plot one
 - Keep|Next
 - Next
 - Prev
 - Keep|Prev
 - 1
 - 1, IC4
 - RV: 8.98%
 - X tal: 2
 - Y tal: -94
 - Z tal: 27
 - Display:**
 - Mesh on
 - Tight view
 - Sagittal view
 - Coronal view
 - Top view
 - No controls**



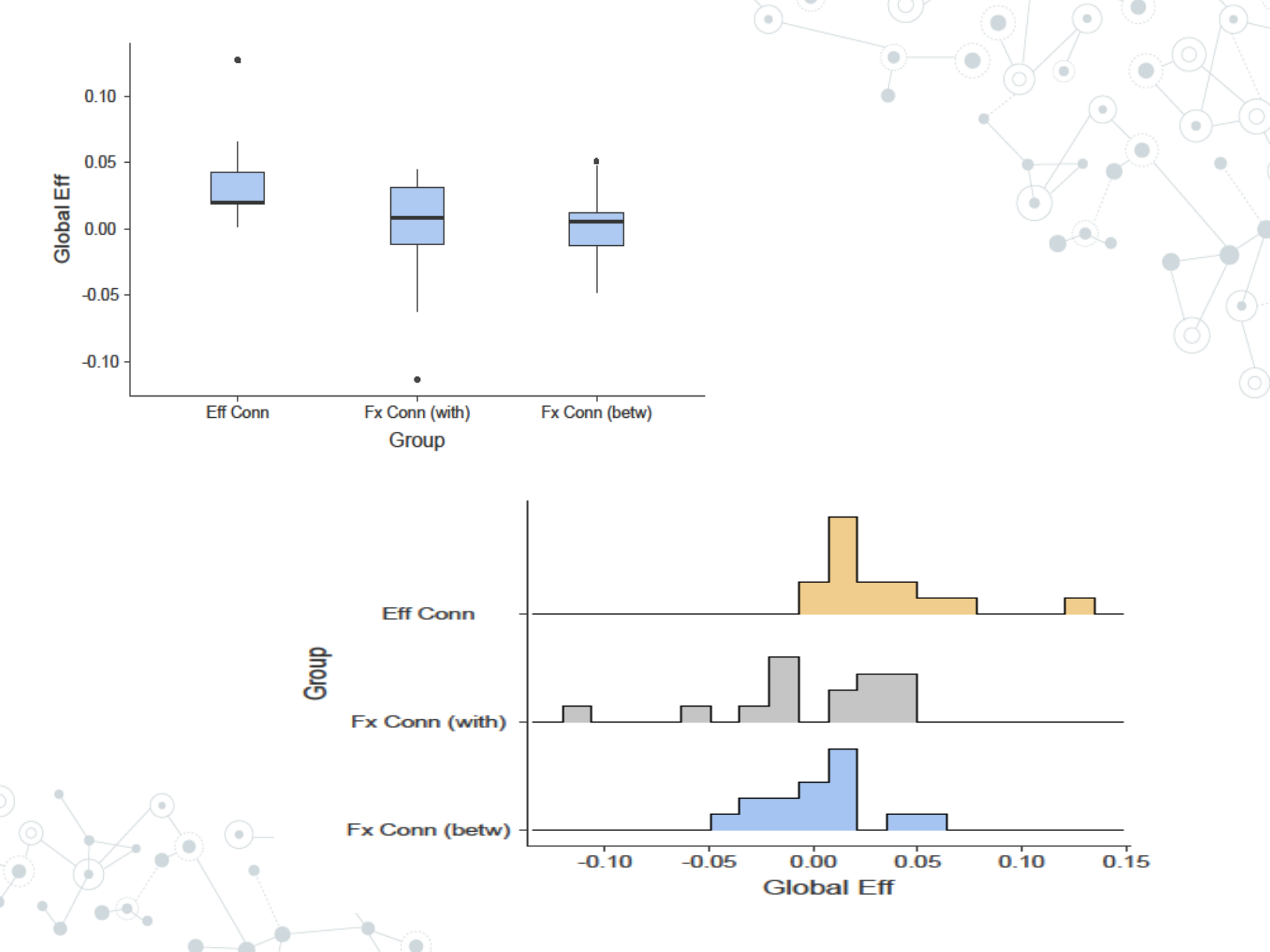
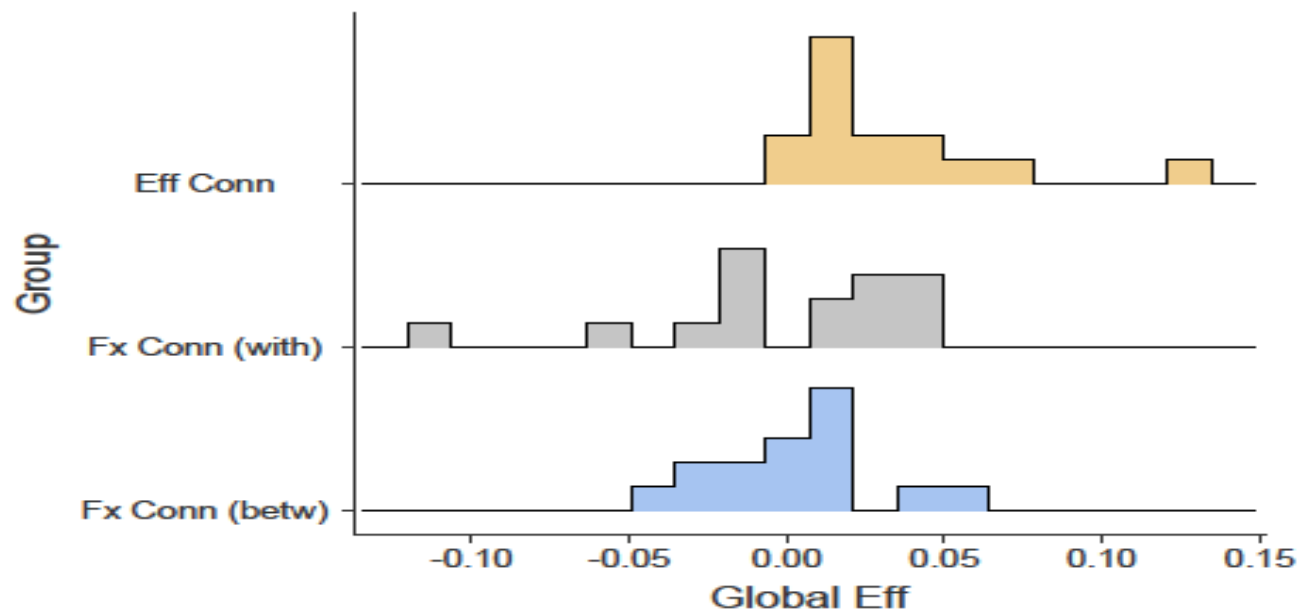
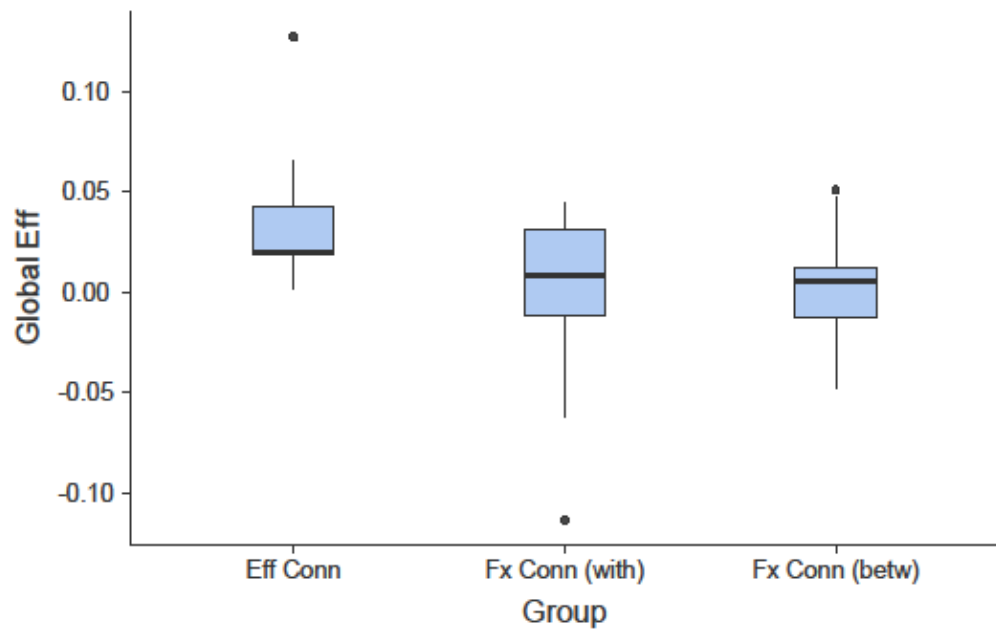


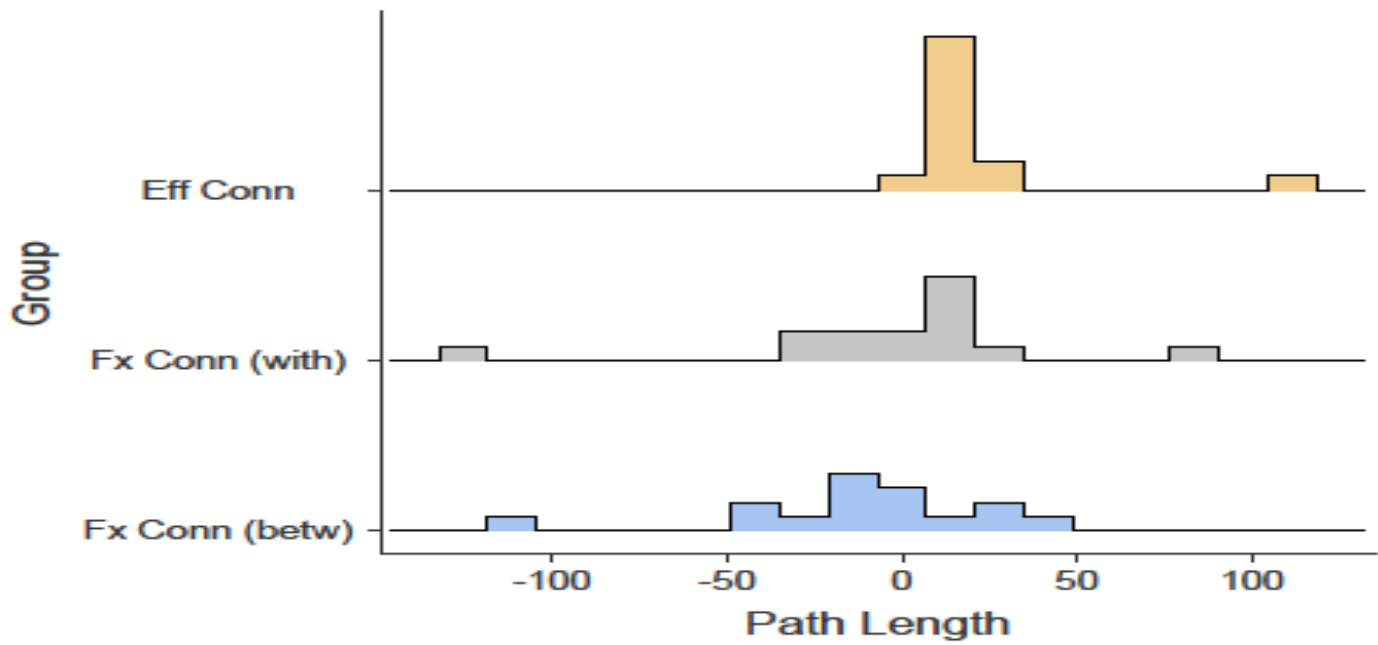
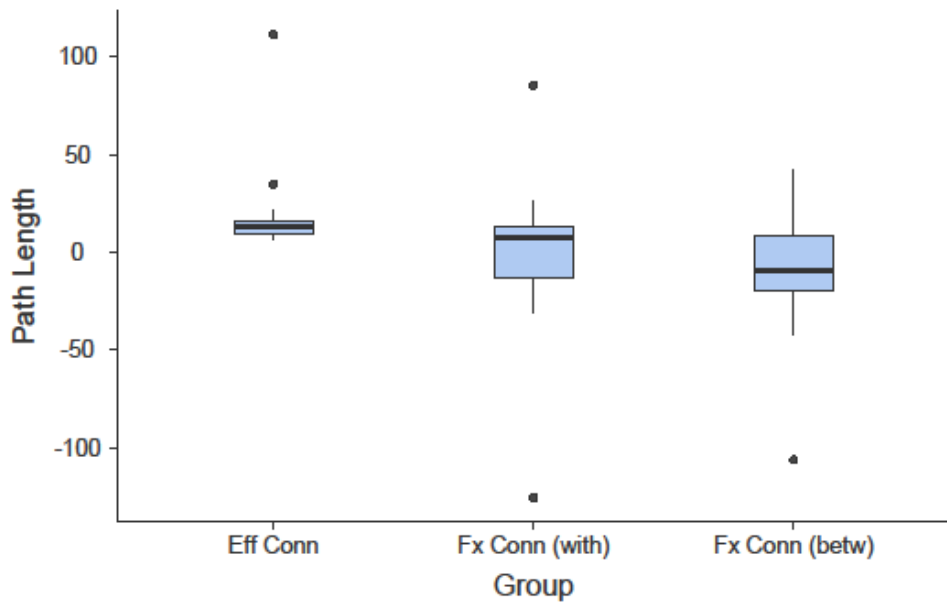


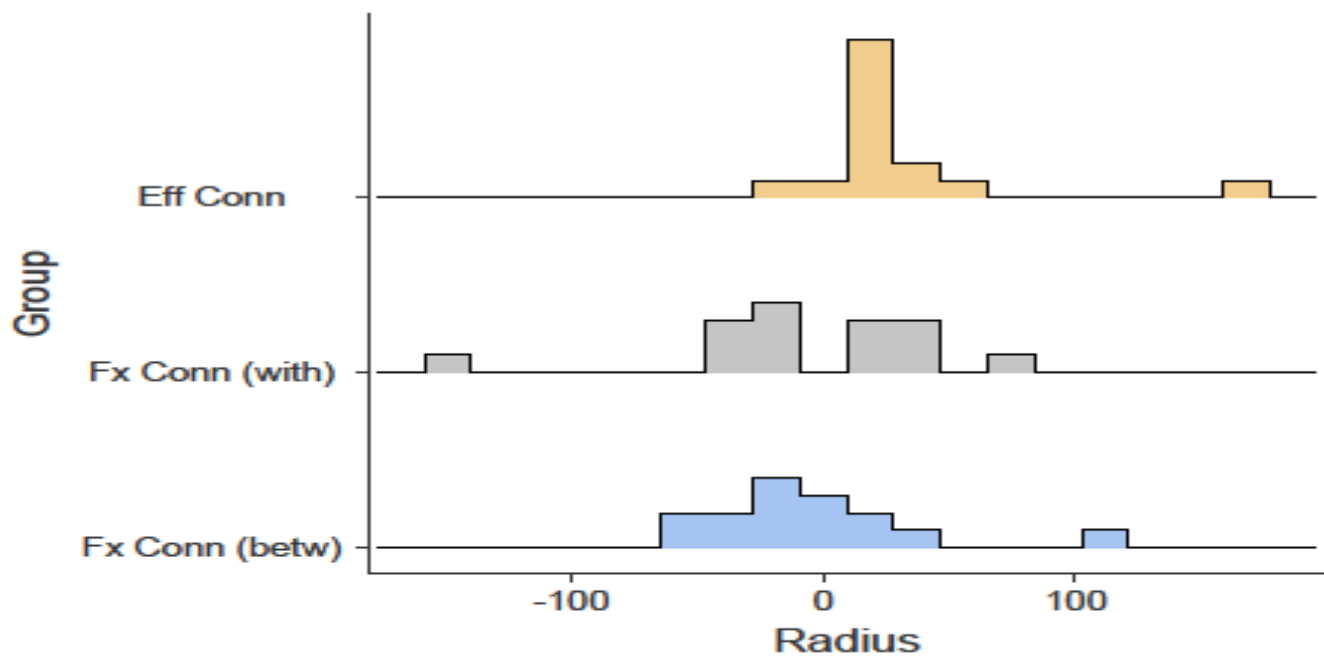
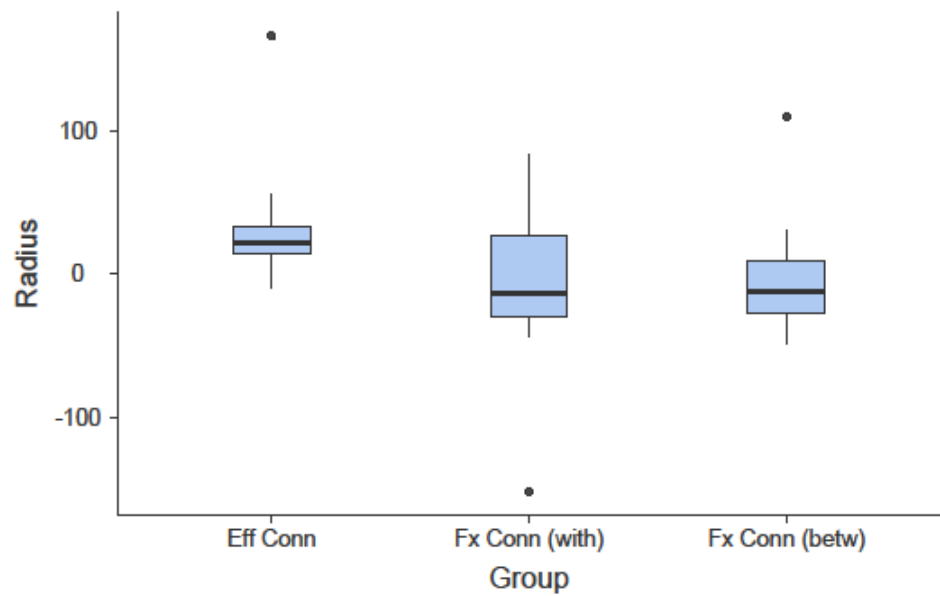
Statistical Analyses of Graph Theory Metrics (Connectivity)

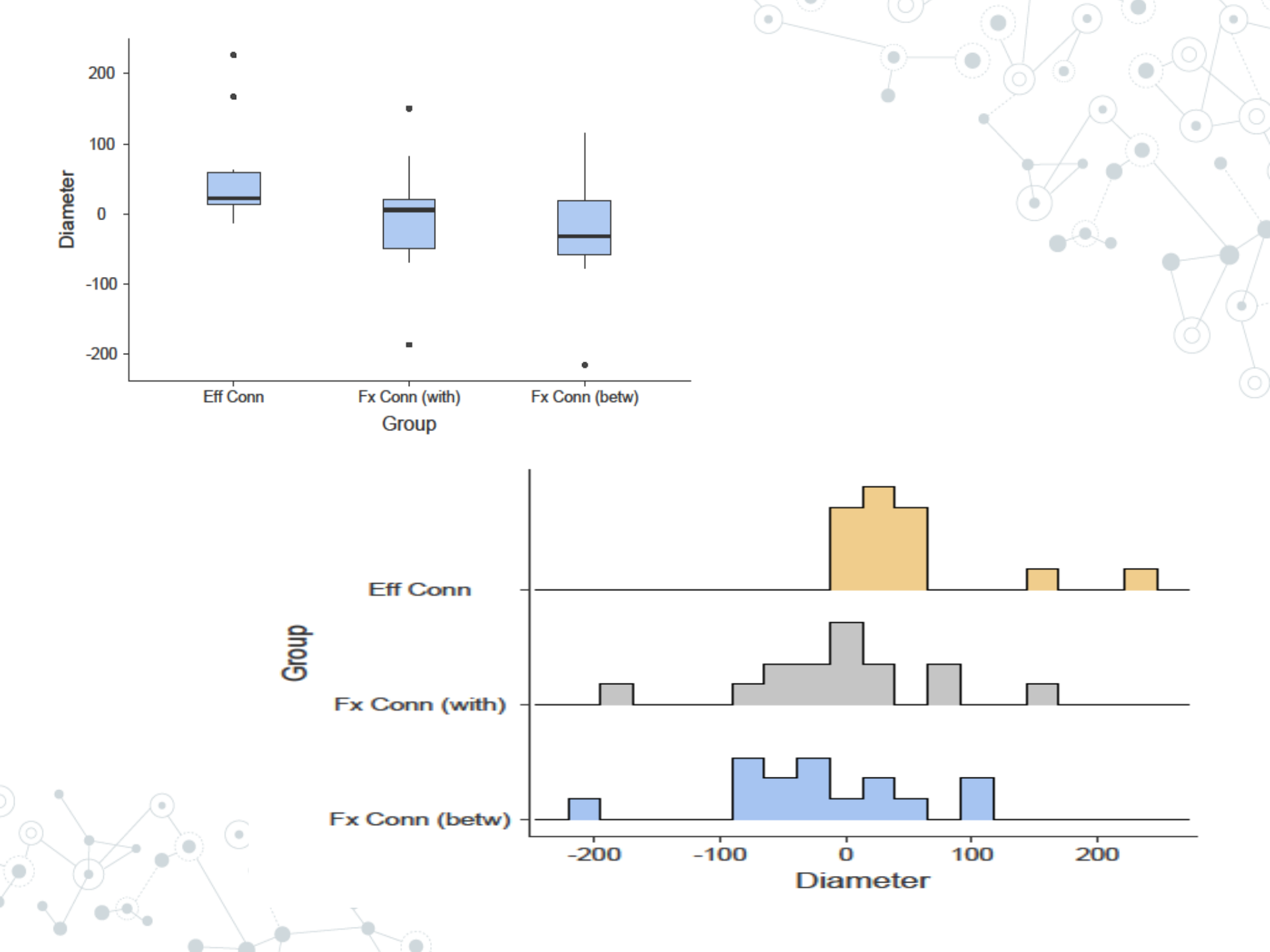
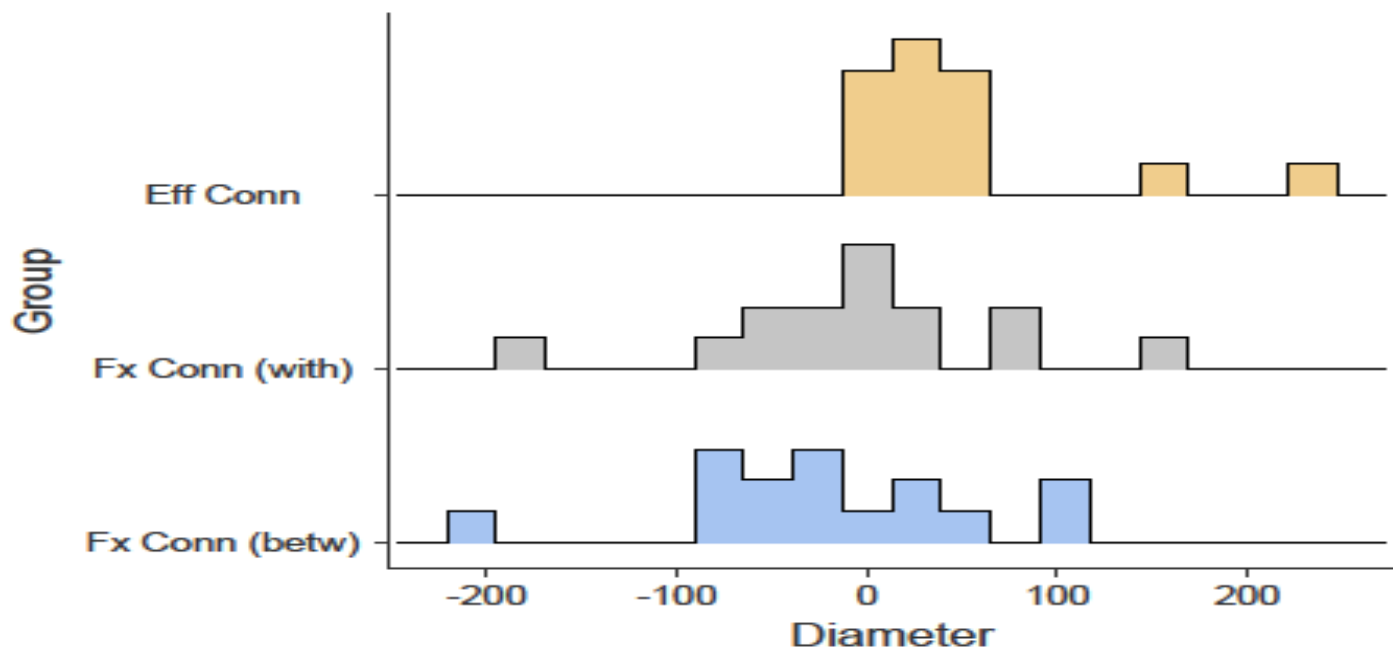
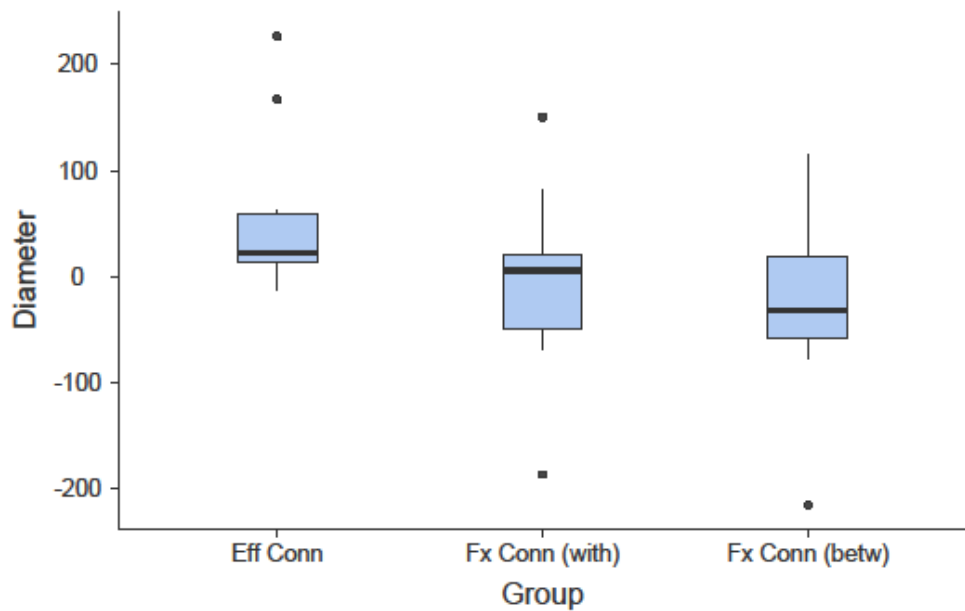
Analysis of Variance

Cluster Coefficient	Global Efficiency	Path Length	Radius	Diameter
F = 0.429	F = 4.60	F = 2.93	F = 3.35	F = 3.70
p = 0.654	p = 0.016	p = 0.064	p = 0.045	p = 0.033









Correlation Matrix

		Medications	Clinical Improvement	Therapist Outcome Measure	Diameter
Medications	Pearson's r	—	-0.142	-0.142	-0.027
	p-value	—	0.353	0.351	0.862
Clinical Improvement	Pearson's r		—	0.563	0.319
	p-value		—	< .001	0.033
Therapist Outcome Measure	Pearson's r			—	0.242
	p-value			—	0.109
Diameter	Pearson's r				—
	p-value				—

Conclusions

- ◎ Measures of effective connectivity can be gleaned from QEEG data.
- ◎ Effective connectivity guided multivariate coherence training led to enhanced client and therapist ratings of outcome.
- ◎ Therapist ratings are consistently higher than clients and show more significant differences.
- ◎ Both ratings show an increased likelihood of greater outcomes (> 10) in the effective connectivity group.
- ◎ Positive NF outcomes in this group showed greater reductions of delta/theta, alpha and beta frequencies. These were commonly seen over bilateral posterior brain regions including temporal locations and midline frontal locations as well.
- ◎ Positive NF outcomes were associated with greater changes in multivariate connectivity. This is especially true for long range connectivity (diameter).
- ◎ Use of effective connectivity leads to changes in connectivity and is more likely to prevent negative connectivity changes.

The background of the slide is a light blue-grey color with a complex network pattern. It consists of numerous small, light-colored circles connected by thin, light-colored lines, creating a web-like structure that fills the entire frame. The text is centered and has a subtle reflection effect below it.

Thanks to our team:

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Tarik Bel-Bahar, PhD