



Does Brain Functional Connectivity Alter Across Similar Trials During Imaging Experiments?

Li Zhu, and Laleh Najafizadeh

Integrated Circuits and Neurolmaging Laboratory

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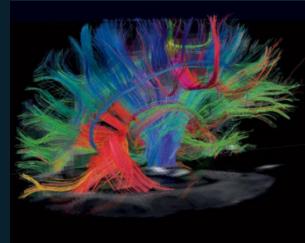
Outline

- **Introduction**
 - Brain Connectivity
 - Motivation
 - Problem Definition
- **Experiment Setup**
 - Task Paradigm
 - Imaging Technique
- **Processing**
 - Preprocessing
 - Similarity Analysis
 - Statistical Testing
- **Results**
- **Conclusion**

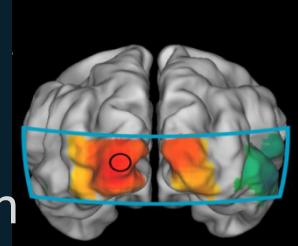


Brain Connectivity

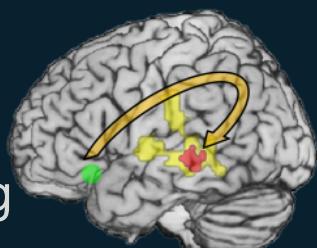
- **Investigates the Networks within the Brain**
- **Anatomical Connectivity**
 - looks for axonal connections
 - diffuse tensor imaging, tracing techniques



- **Functional Connectivity**
 - looks for statistical similarities between regional time series
 - functional neuroimaging techniques, seed-based correlation



- **Effective Connectivity**
 - looks for causal influences between regions of brain
 - functional imaging techniques, causal interactions modeling



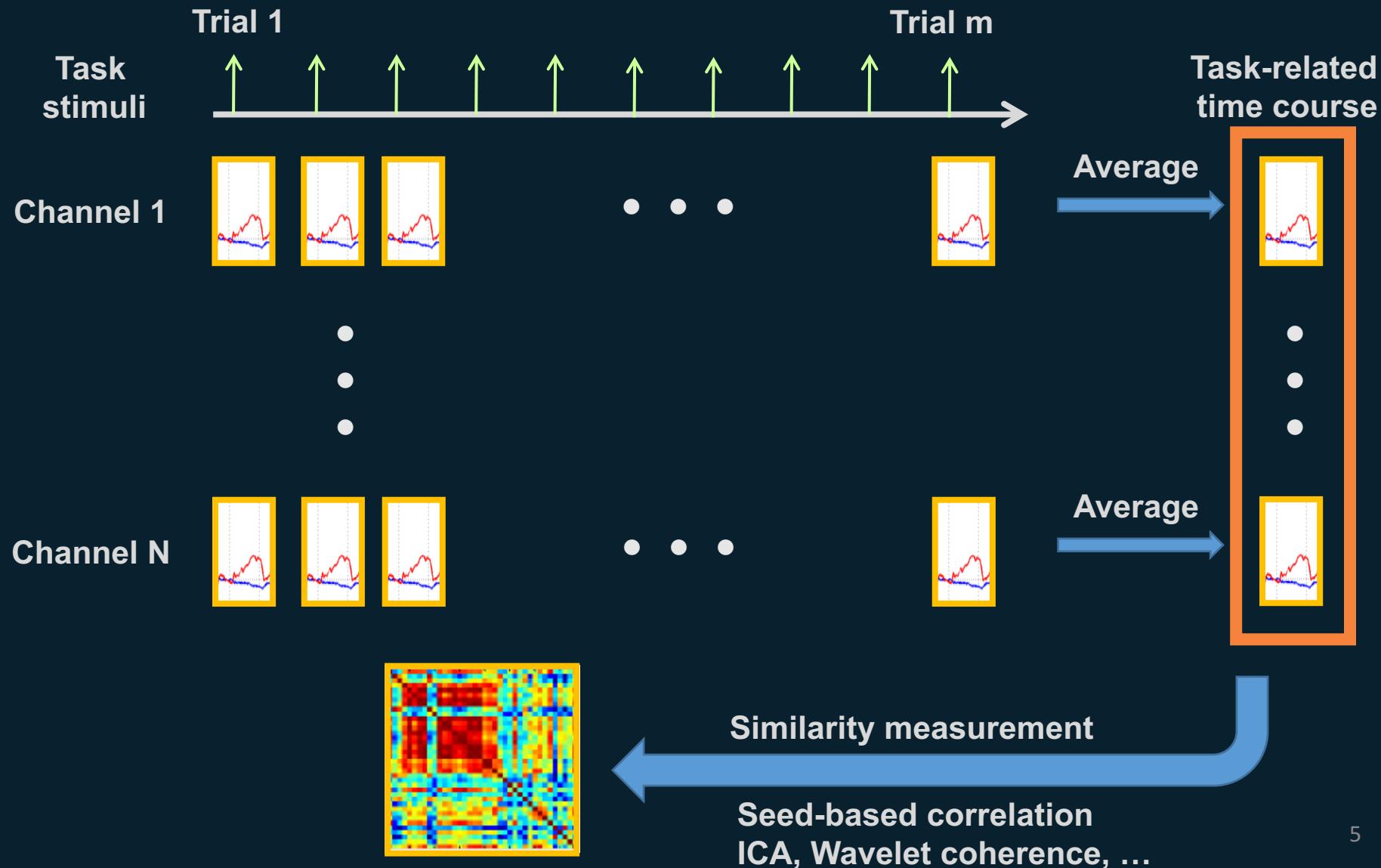


Motivation

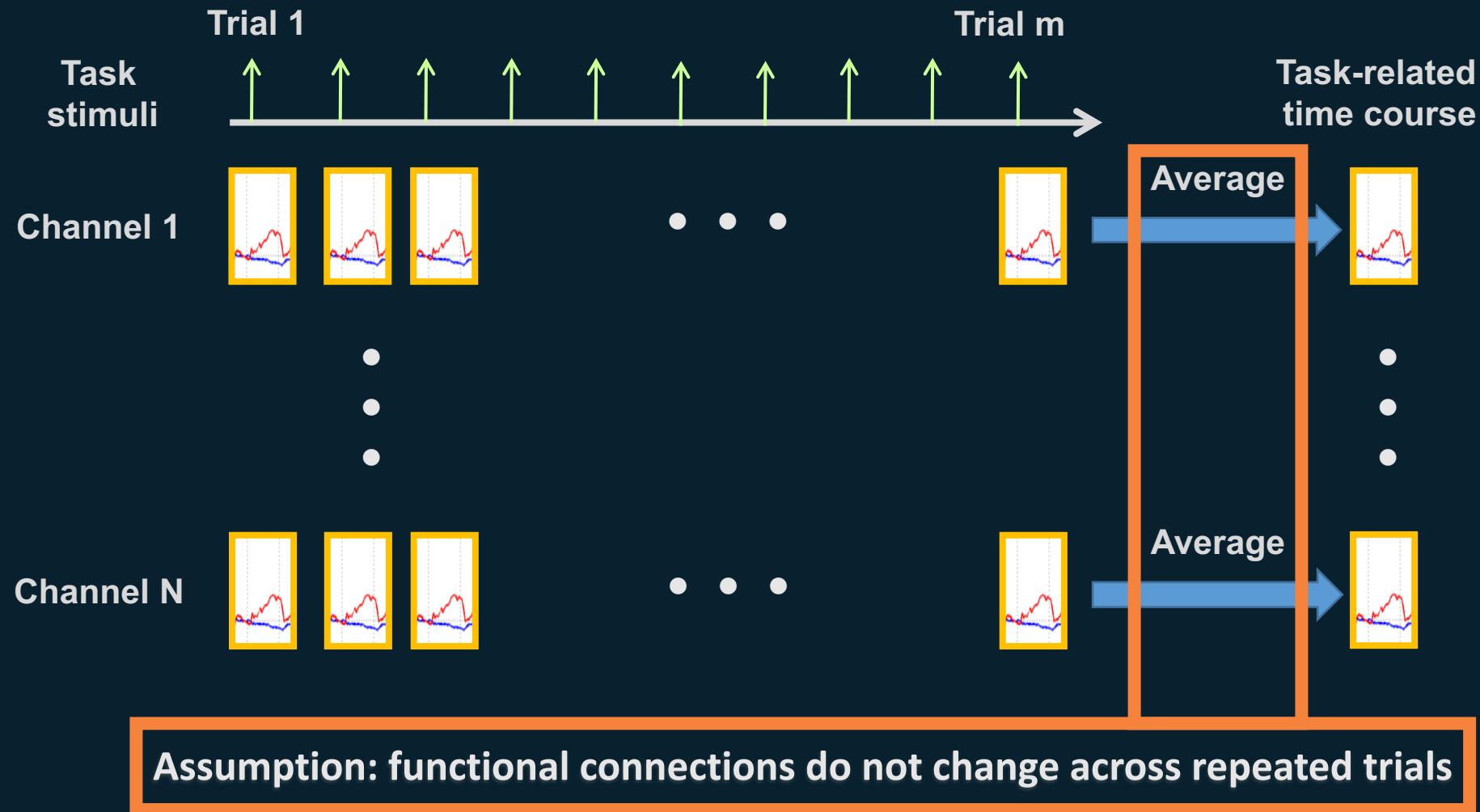
- Understanding the functionality of the brain at the network level
- Potential for early diagnosis of brain-related diseases
 - Autistic Spectrum Disorders
 - 1 in 1000 children are diagnosed with Autism
 - Specific cause not known (biological, neurological, environmental...)
 - Early diagnosis could be a key
 - Schizophrenia
 - 1 in 100 US population are diagnosed with Schizophrenia
 - Abnormal function of neural communication that change the functional connectivity, while the anatomical elements remain intact.



Problem Definition



Problem Definition





Problem Definition

Variability in the brain activation from trial to trial has been reported in previous studies:

- L. Holper, N. Kobashi, D. Kiper, F. Scholkmann, M. Wolf, and K. Eng, “Trial-to-trial variability differentiates motor imagery during observation between low versus high responders: a functional near-infrared spectroscopy study.” *Behavioural brain research*, vol. 229, 2012, pp. 29–40.
- XS Hu, KS Hong, and SS Ge, “Reduction of trial-to-trial variability in functional near-infrared spectroscopy signals by accounting for resting-state functional connectivity,” *Journal of biomedical optics*, 2013.
- M. Fox, A. Snyder, J. Zacks, and M. Raichle, “Coherent spontaneous activity accounts for trial-to-trial variability in human evoked brain responses,” *Nature Neuroscience*, vol. 9, 2005, pp. 23–25.

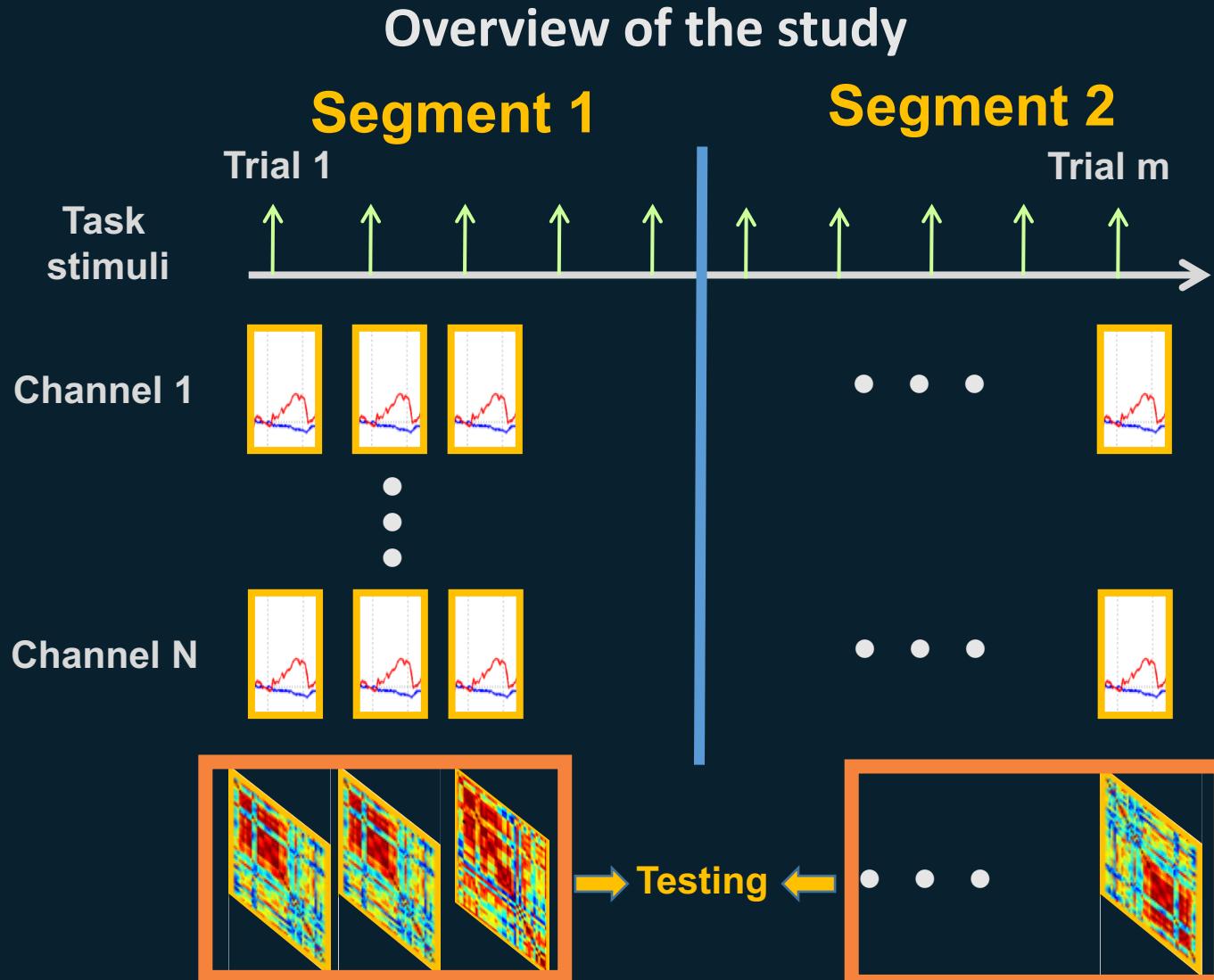


Problem Definition

Functional connections do not change across repeated trials

Is it true?

Problem Definition





Outline

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- **Experiment Setup**
 - Task Paradigm
 - Imaging Technique

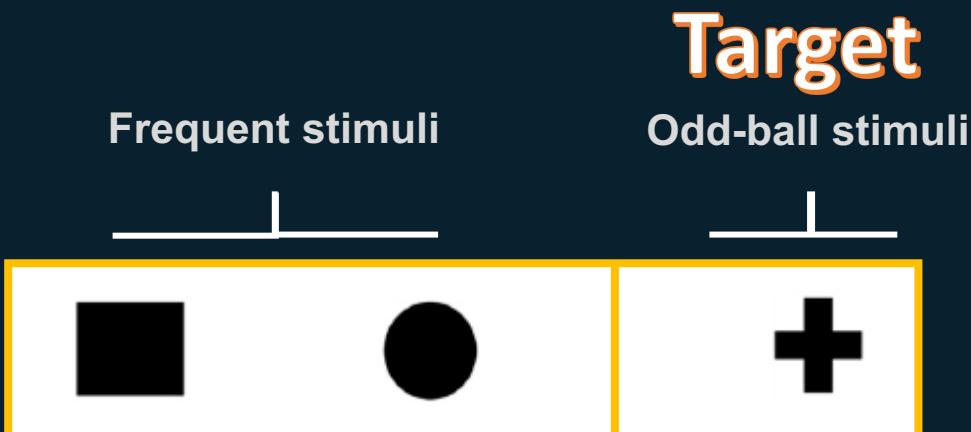
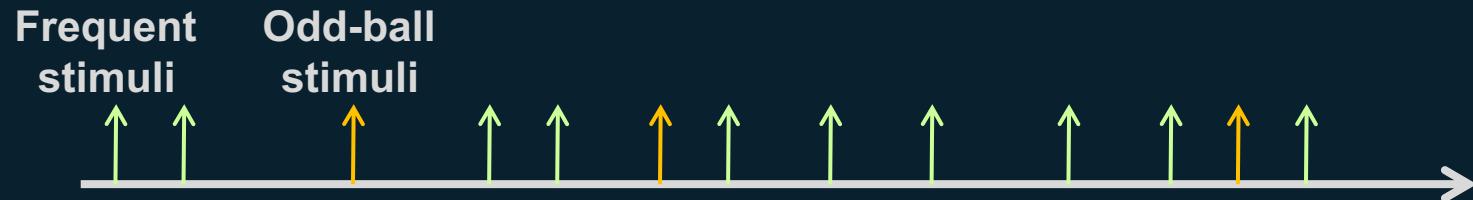
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- **Results**

- **Conclusion**

Task Paradigm

- **Modified Visual Odd Ball Task**



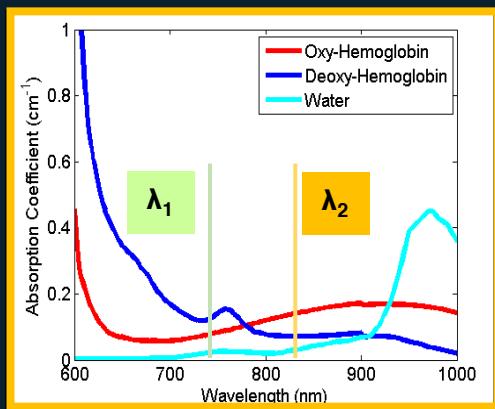
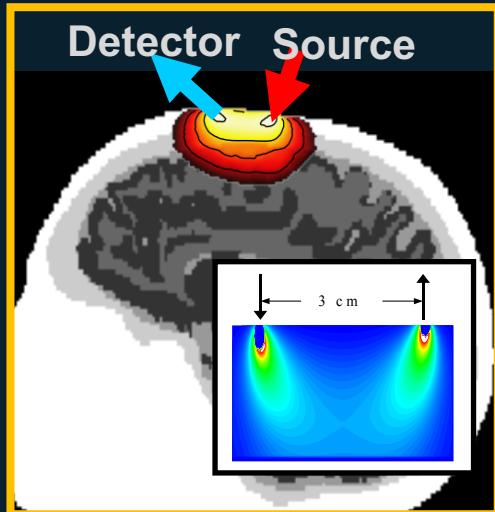
Target

Odd-ball stimuli

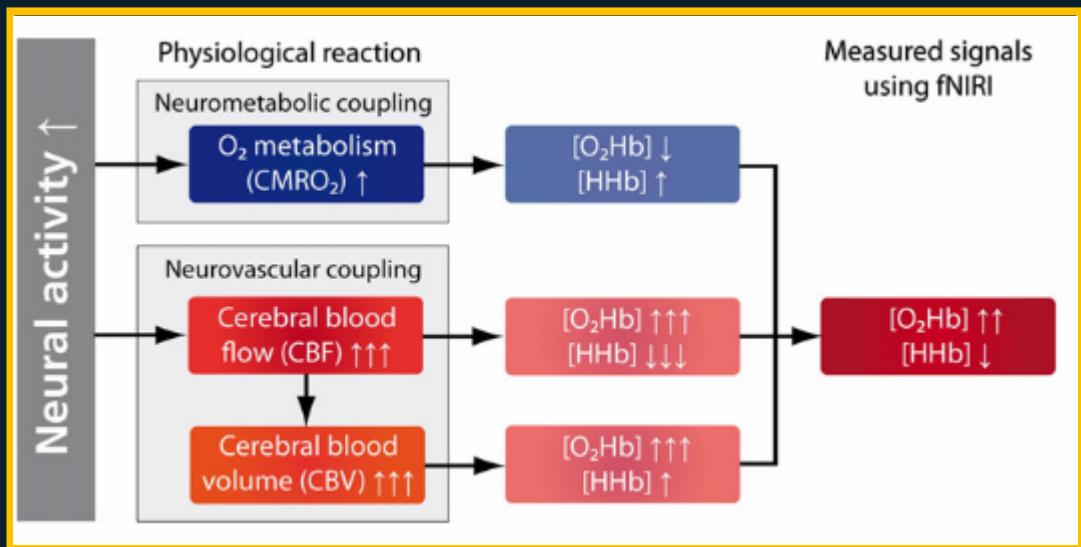
- 30 target trials
- 190 frequent stimuli trials
- Presentation time = 50 ms
- ITI = 10 – 12 s
- Left click if see target
- 5 healthy participants

Imaging Technique

- Functional Near-Infrared Spectroscopy



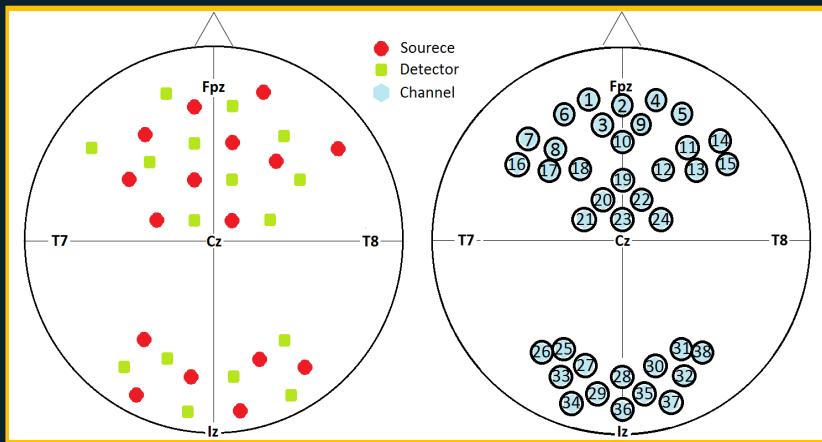
Overview of neurovascular mechanism



F. Scholkmann et. Al., "A review on continuous wave functional near-infrared spectroscopy and imaging instrumentation and methodology," *NeuroImage*, vol. 85 Pt 1, Jan. 2014, pp. 6–27.

Imaging Technique

- Functional Near-Infrared Spectroscopy



- 16 sources, 16 detectors
- 38 channels
- cover prefrontal/visual cortices
- 760 and 830 nm
- Sampling rate: 10.42 Hz
- Spatial Resolution: 3 cm
- Stimuli Sent by E-prime

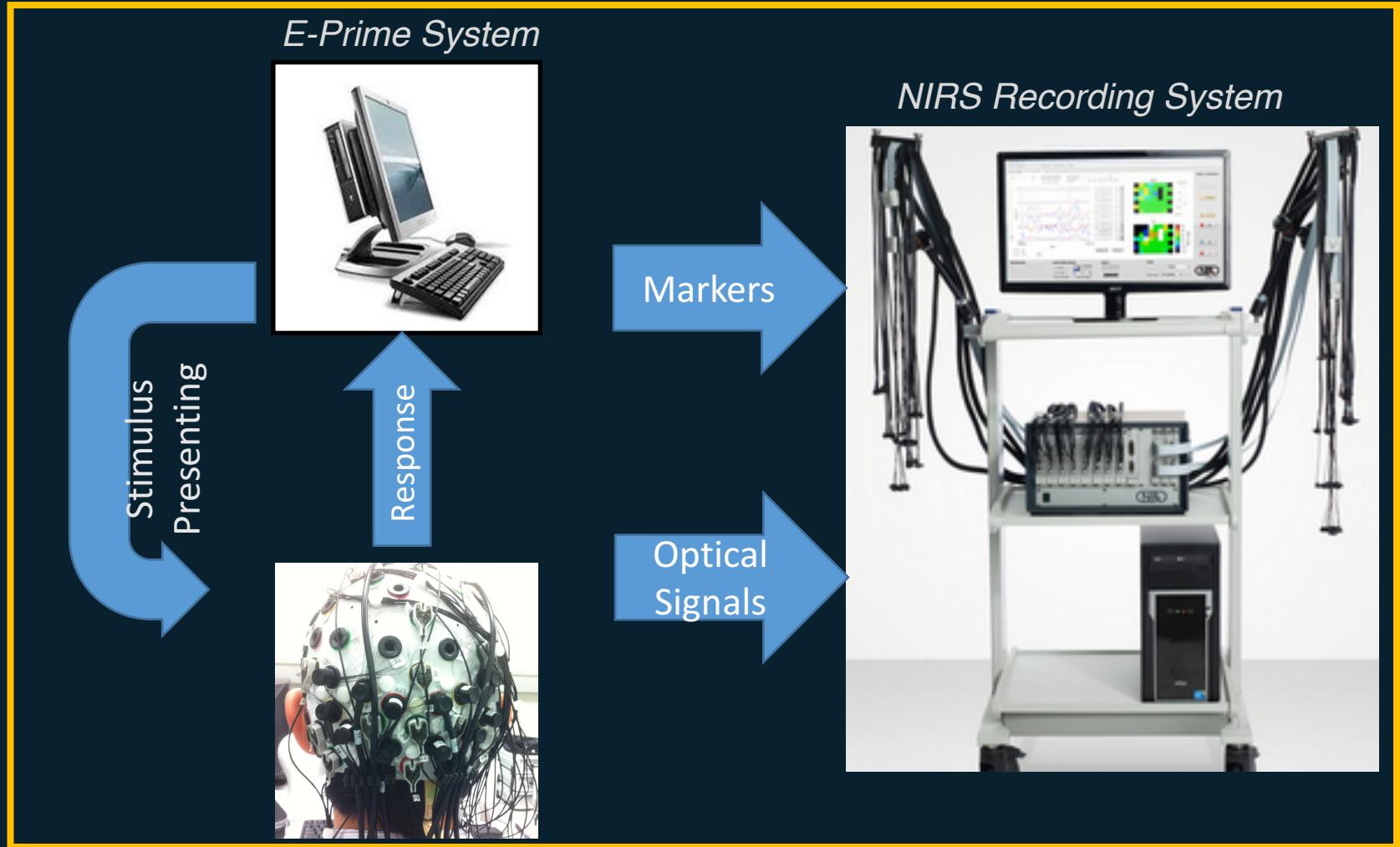
NIRx Medical Technologies
Model: NIRScout



www.nirx.net/imagers/nirscout

Imaging Technique

- Experimental Setup





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Preprocessing

Preprocessing



Wavelet transform coherence



Data organization & testing

- Extracting Brain Activities

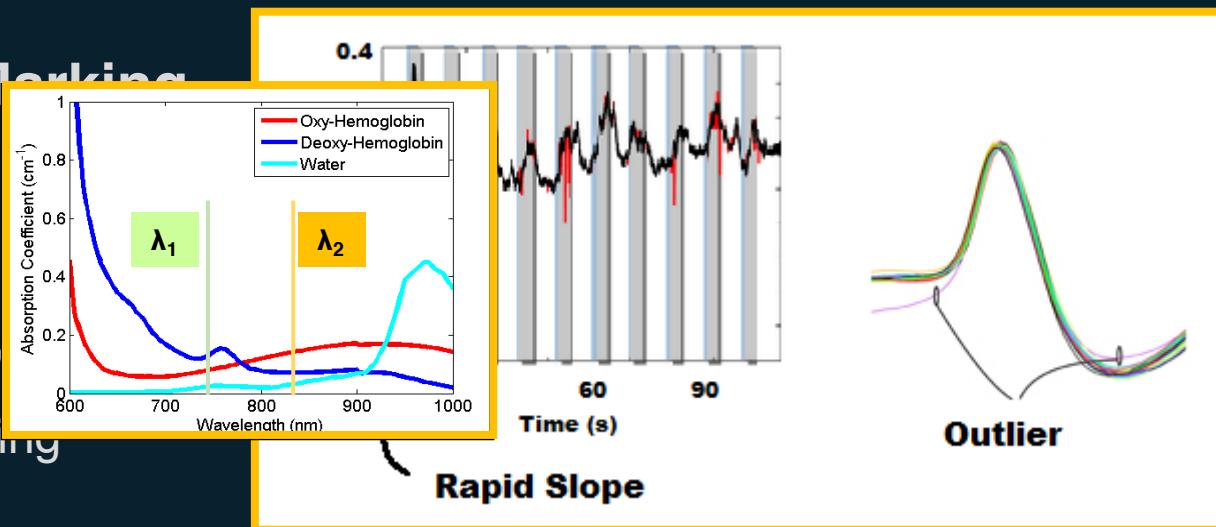
- both ΔHbO_2 and ΔHbR were extracted using Modified Beer Lambert Law:

$$\text{Wavelength 1 (760 nm): } \ln\left(\frac{I_{\text{task},\lambda_1}}{I_{\text{baseline},\lambda_1}}\right) = -(\epsilon_{HbO_2,\lambda_1} \Delta C_{HbO_2} + \epsilon_{HbR,\lambda_1} \Delta C_{HbR}) \cdot L_{\lambda_1}$$

$$\text{Wavelength 2 (830 nm): } \ln\left(\frac{I_{\text{task},\lambda_2}}{I_{\text{baseline},\lambda_2}}\right) = -(\epsilon_{HbO_2,\lambda_2} \Delta C_{HbO_2} + \epsilon_{HbR,\lambda_2} \Delta C_{HbR}) \cdot L_{\lambda_2}$$

- Artifacts Detection/Marking

- Detrending
- Segmentation
- Baseline Correction
- Rapid slope Detection/
- Outlier Detection/Marking



Wavelet Transform Coherence

Preprocessing

Wavelet transform
coherence

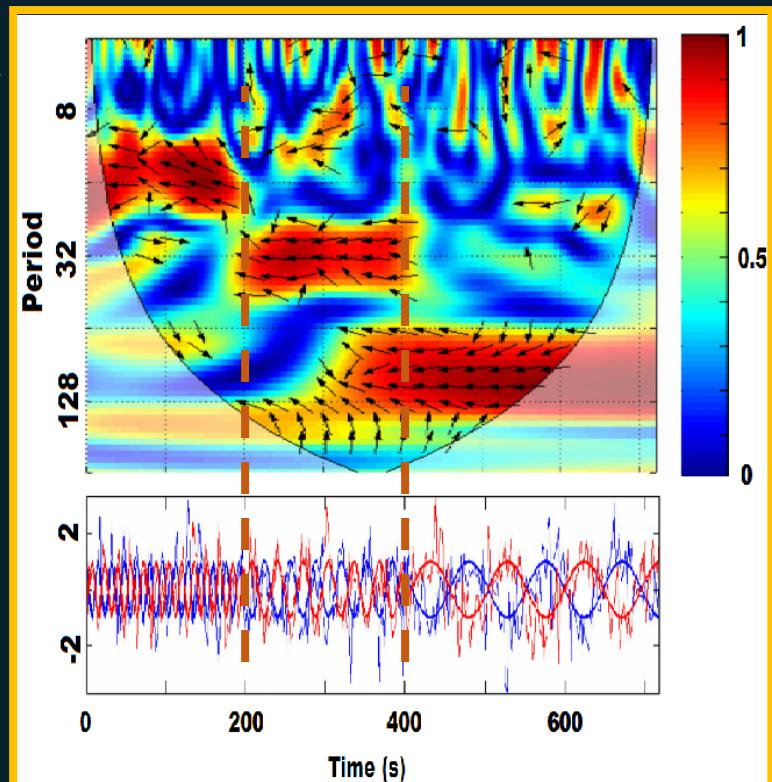
Data organization
& testing

- Wavelet transform used for investigating the time-frequency features of the non-stationary signals:

$$W_{x_n}(n, s) = \sqrt{\frac{\Delta t}{s}} \sum_{m=1}^N x_m \Psi_0^*[(m - n) \frac{\Delta t}{s}]$$

- Wavelet transform coherence measures the cross-correlation between two time series in both time and frequency domain:

$$R_{x_n, y_n}^2(n, s) = \frac{|Smooth(s^{-1}W_{x_n, y_n}(n, s))|^2}{Smooth(s^{-1}|W_{x_n}(n, s)|^2) \cdot Smooth(s^{-1}|W_{y_n}(n, s)|^2)}$$



Simulated signals

Wavelet Transform Coherence

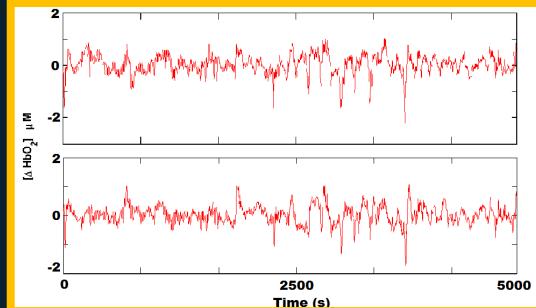
Preprocessing

Wavelet transform
coherence

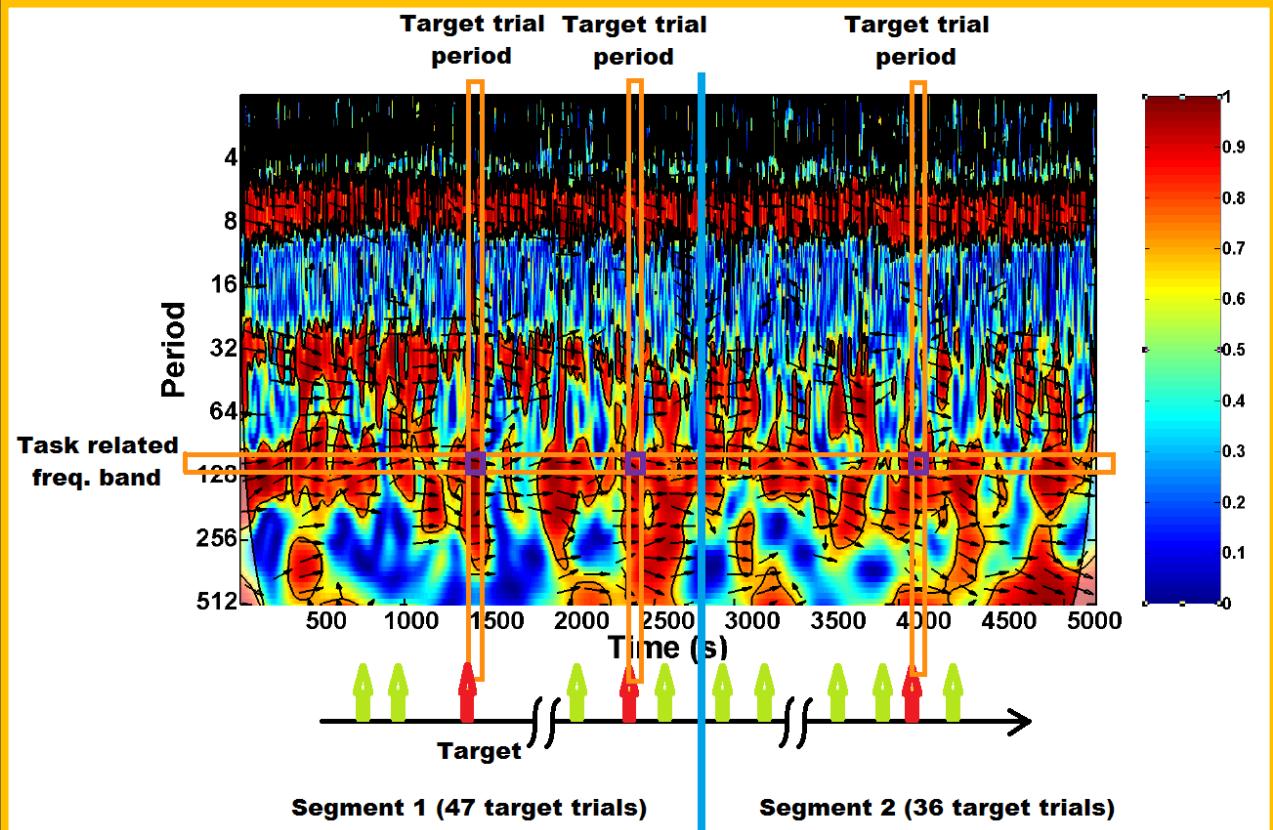
Data organization
& testing

- WTC was computed for each channel-pair.

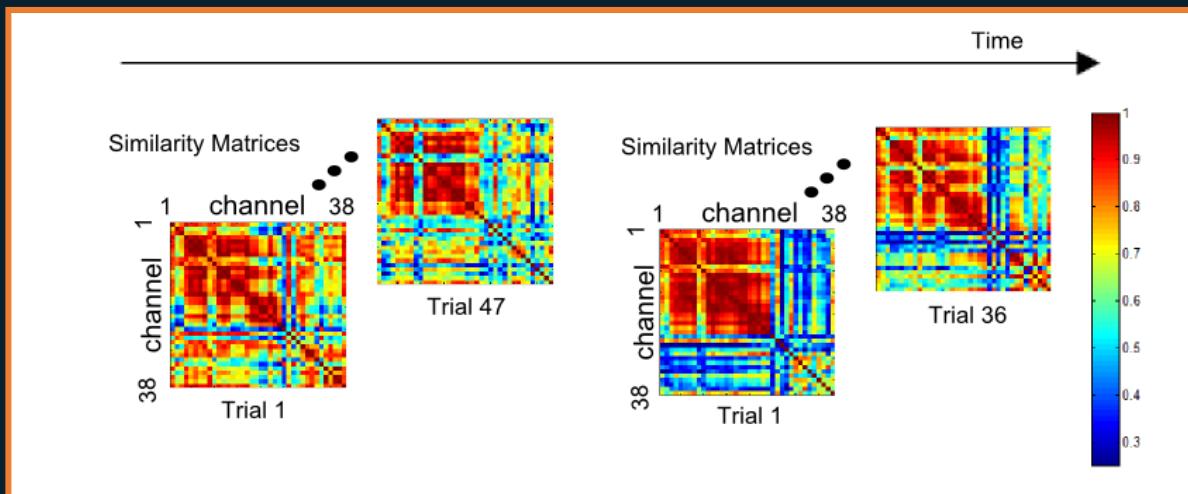
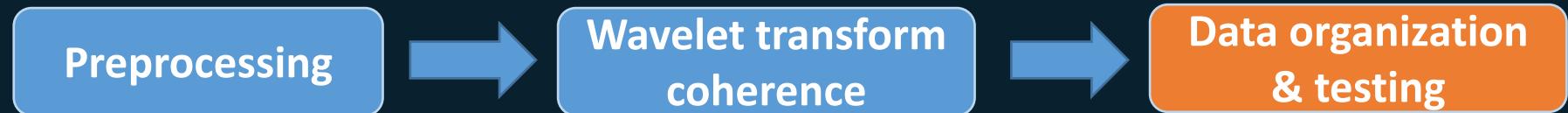
Channel 1 vs Channel 26



WTC
Computing



Data organization & testing



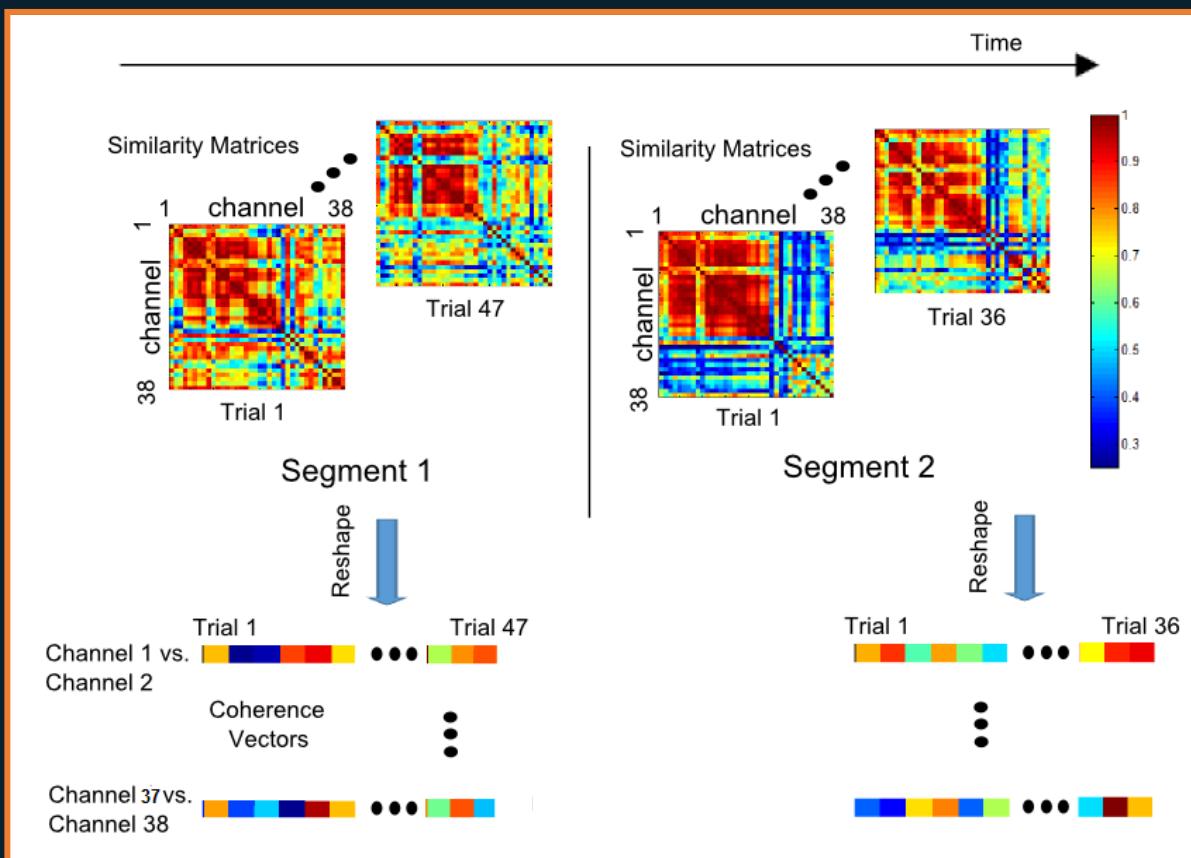
- The similarity measurements were organized into a series of matrices.

Data organization & testing

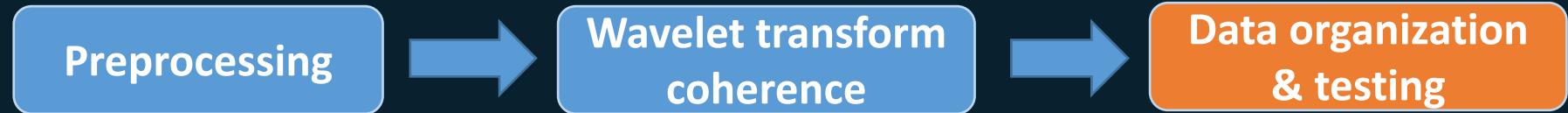
Preprocessing

Wavelet transform
coherence

Data organization
& testing

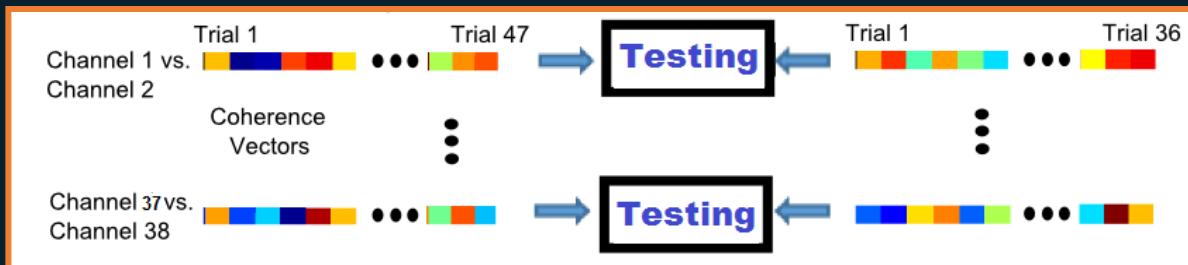


Data organization & testing

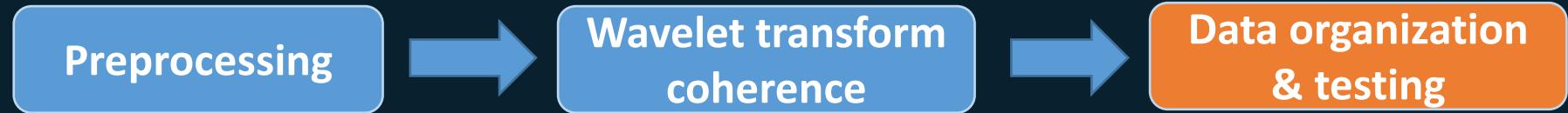


■ Statistical Testing

- Statistical testing was implemented for each channel-pair.
- Null hypothesis H_0 : The measure of functional connectivity from two segments are equivalent.
- Non-parametric permutation testing: complete freedom with respect to the distribution of the data or their parameters, which could increase the sensitivity of the statistical test.



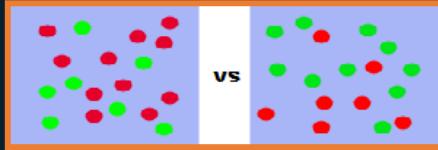
Data organization & testing



■ Permutation testing

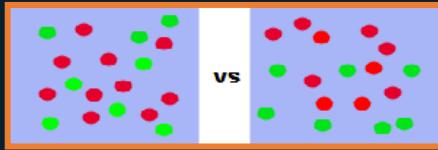
- No need distributional assumptions
- If H_0 is true, shuffling the data won't affect the test statistics
- Algorithm
 - Shuffle data across segments, compute t-value
 - Repeat 1000 times, obtain null distribution
 - Compute t-value t_0 from the observed datasets
 - Compute p-value using t_0 with the null distribution

Shuffle
1



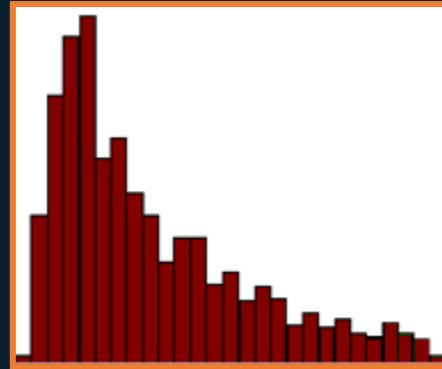
⋮

Shuffle
1000

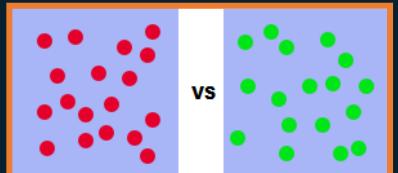


Wavelet transform
coherence

Data organization
& testing

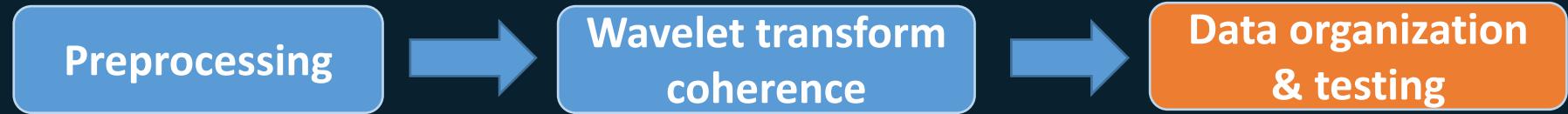


Original
Data



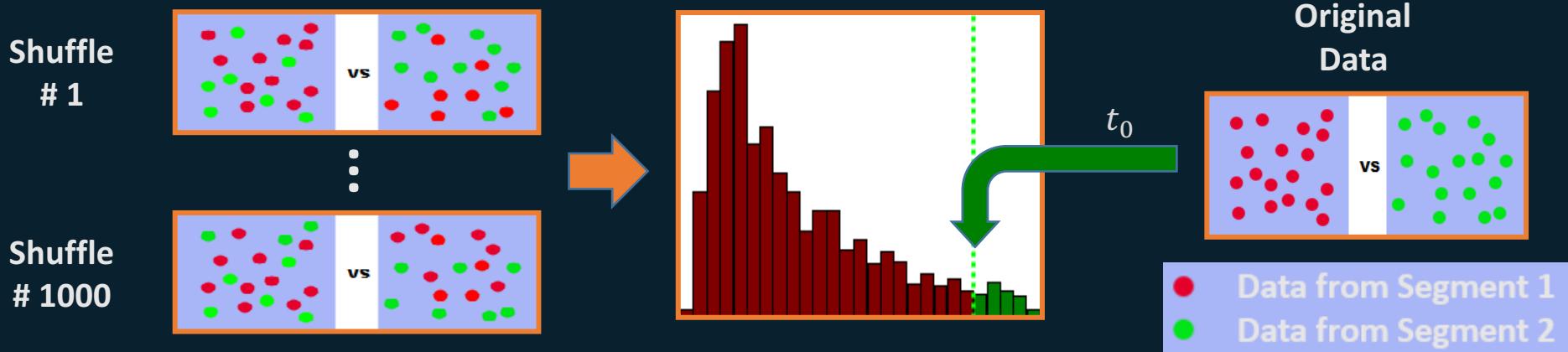
● Data from Segment 1
● Data from Segment 2

Data organization & testing



■ Permutation testing

- No need distributional assumptions
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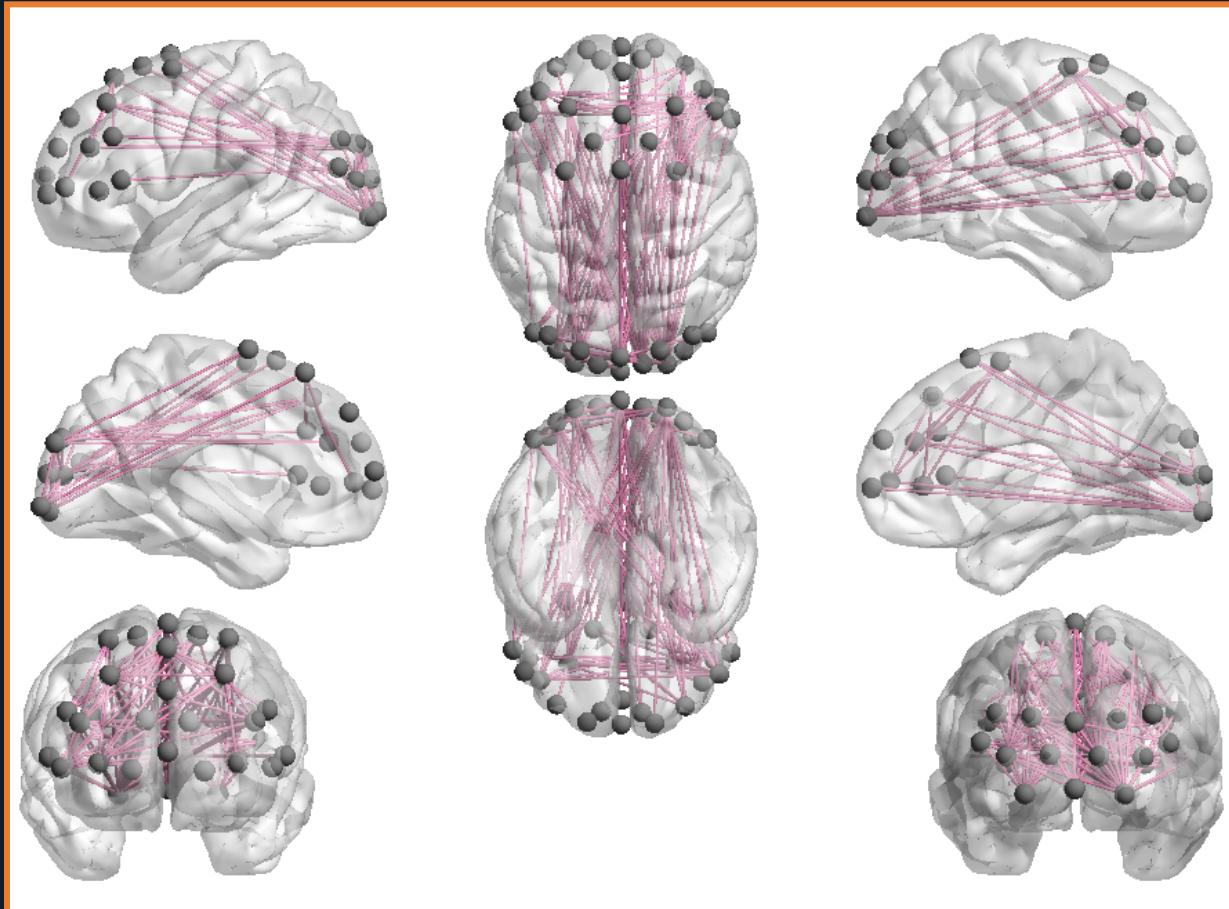
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Results

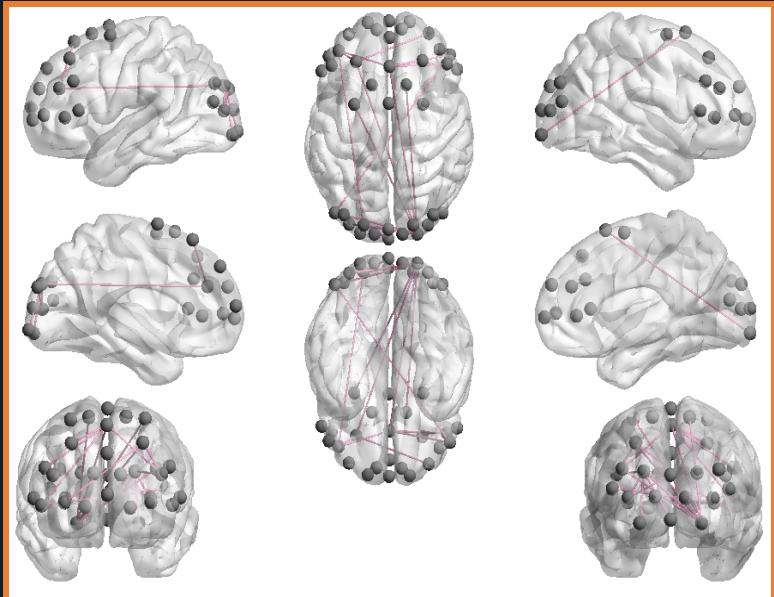
- Channel-pairs with Significant variation ($p<0.01$)
Segment 1 vs Segment 2



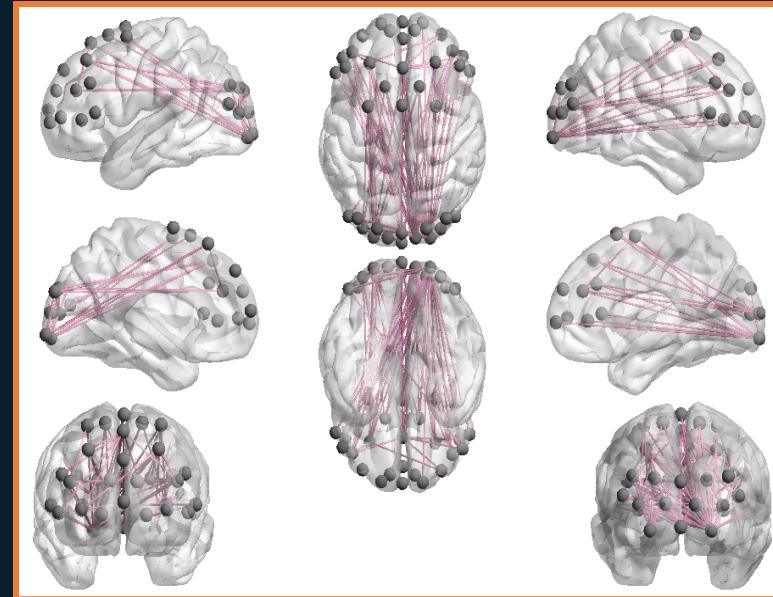
Results

- Channel-pairs with Significant variation ($p<0.05$)

Segment 1 vs
whole number of trials



Segment 2 vs
whole number of trials





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Conclusion and Future Work

- There exists channel-pairs that revealed significant variation between the two temporal segments in their functional connections
- Extracting the characteristics of the brain networks from all similar trials in a long-lasting experiment might introduce bias in the result
- The analysis procedure in this study might be extended to be used for determining the maximum number of similar trials in an experiment for which functional connections between brain networks do not vary significantly



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Thank you!