

Workshop on Microlocal Methods in Medical Imaging

# A New Adaptive S Transform with Applications in Studying Brain Functions

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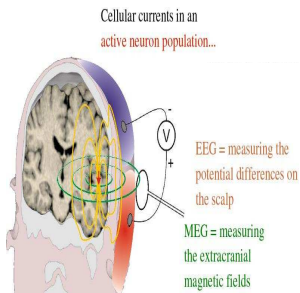


# Outline

- 1 Introduction
- 2 An Adaptive Multi-resolution Time-frequency Representation
- 3 Time-varying Spectral Measures for Analyzing Brain Time Series
- 4 Studying Motor Activities using the Multi-source Interference Task
- 5 Summary and Future Work

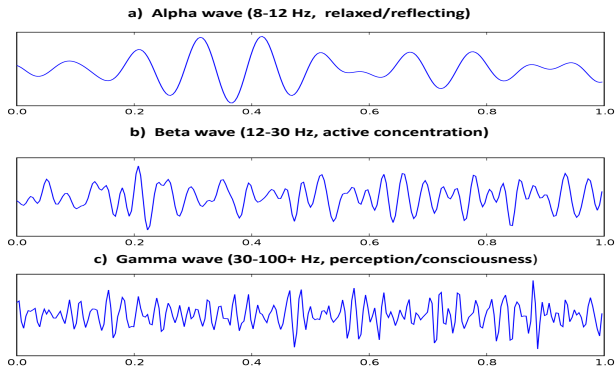
# Studying brain functions

- Understand the brain's control mechanisms
- Guide treatment of mental disorders and other neurological diseases
- Brain signals:  
Electroencephalography (EEG) and  
Magnetoencephalography (MEG)



Adapted from Lauri Parkkinen, HBM course 2006

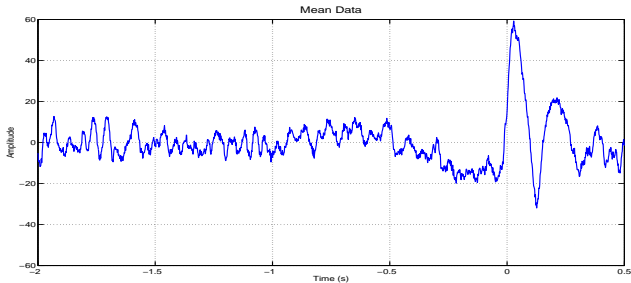
# Brain rhythms



- Spectrum analysis becomes a popular approach in analyzing brain signals.

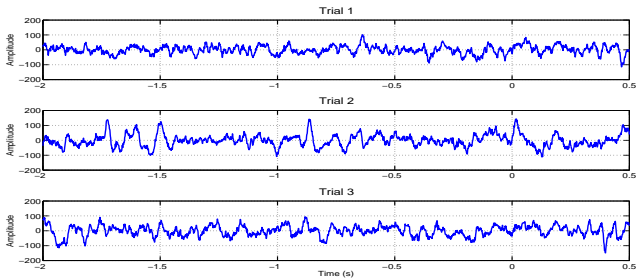
# Challenges in brain research

- Non-stationarity
- A huge range of variations of the signal structure
- Complication of interactions among different brain regions



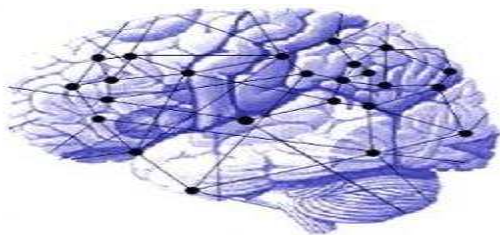
# Challenges in brain research

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# Challenges in brain research

- Non-stationarity
- A huge range of variations of the signal structure
- **Complication of interactions among different brain regions**



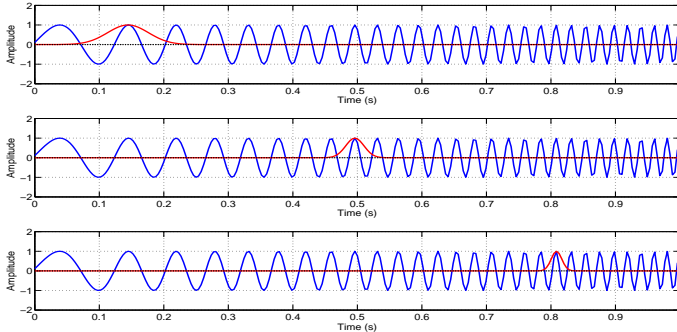
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# The Stockwell transform

- A moving window Fourier transform whose window function is scaled by  $\frac{1}{|f|}$ .



# A generalized Stockwell transform (GST)

- We propose a generalized Stockwell transform with the window function scaled by  $\sigma(f) = \frac{p}{|f|^q}$ . (Pinnegar and Mansinha, 2003:  $\sigma(f) = \frac{p}{|f|}$ ; Sejdić, et al., 2008:  $\sigma(f) = \frac{1}{|f|^q}$ )

## Definition: The Generalized Stockwell Transform

Let  $\psi \in L^2(\mathcal{R})$ . The generalized Stockwell transform of  $x \in L^2(\mathcal{R})$  is defined as

$$\text{GST}_x^{(p,q)}(t, f) = \frac{|f|^q}{p} \int_{-\infty}^{\infty} x(\tau) \overline{\psi\left(\frac{|f|^q(\tau - t)}{p}\right)} e^{-j2\pi\tau f} d\tau, \quad f \neq 0,$$

and

$$\text{GST}_x^{(p,q)}(t, 0) = \int_{-\infty}^{\infty} x(\tau) d\tau,$$

where  $p > 0$  and  $q \geq 0$ .

# Main mathematical properties

## A Frequency Domain Equivalence

$$\text{GST}_x^{(p,q)}(t, f) = \int_{-\infty}^{\infty} X(\alpha) \overline{\Psi\left(\frac{p(\alpha - f)}{|f|^q}\right)} e^{j2\pi(\alpha - f)t} d\alpha, \quad f \neq 0.$$

## Connection to the Fourier Spectrum

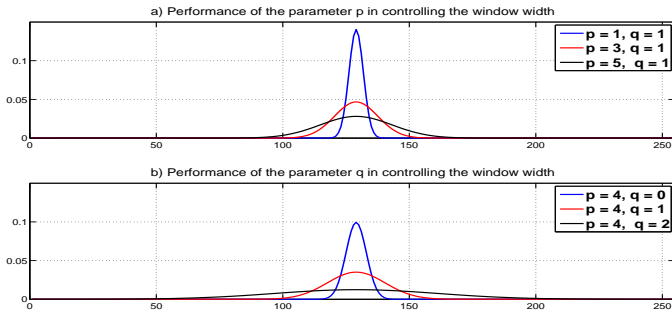
$$\int_{-\infty}^{\infty} \text{GST}_x^{(p,q)}(t, f) dt = X(f).$$

## An Inversion Formula

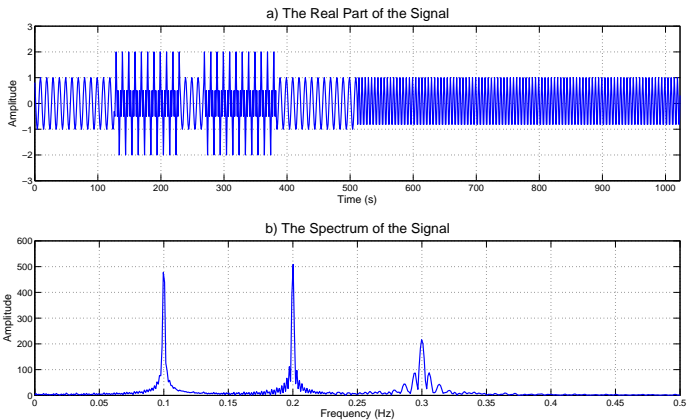
$$x(t) = \mathcal{F}^{-1} \left\{ \int_{-\infty}^{\infty} \text{GST}_x^{(p,q)}(t, f) df \right\}.$$

# Effects of the parameters $p$ and $q$ ( $\sigma(f) = \frac{p}{|f|^q}$ )

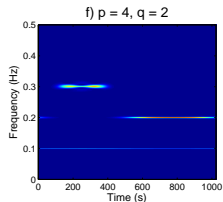
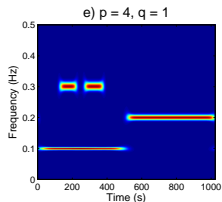
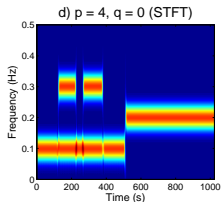
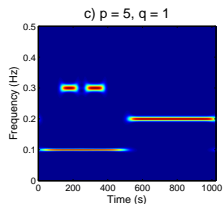
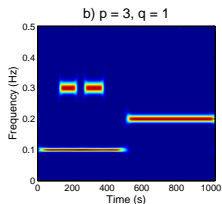
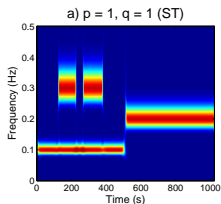
- The width of the window function in the GST increases as the value of  $p$  or  $q$  increases, and  $q$  has a stronger influence.



# A simulated signal



# An illustration ( $\sigma(f) = \frac{p}{|f|^q}$ )



# Question

- Given a specific signal, how to automatically identify optimal parameters such that the corresponding GST provides better resolution to reveal the signal characteristics?

# Evaluate the GST performance by measuring the energy concentration

- A good performed time-frequency representation (TFR) is expected to have the signal energy only concentrated at the involved frequencies.
- Energy concentration measures have been widely used to evaluate the performance of the TFR.
- The Stanković's measure (2001) with an order  $k$ :

$$CM_k(\text{TFR}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |\text{TFR}_x(t, f)|^{\frac{1}{k}} dt df,$$

where  $\text{TFR}_x(t, f)$  is an energy distribution, *i.e.*,  
 $\int \int_{-\infty}^{\infty} \text{TFR}_x(t, f) dt df = \|x\|_2^2$ .



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# The energy distribution based on the $\text{GST}^{(p,1)}$

## Theorem: Resolution of the Identity Formula of the $\text{GST}^{(p,1)}$

Let  $\psi \in L^2(\mathcal{R})$  be such that

$$C_\psi^p = \int_0^\infty \frac{|\Psi(f-p)|^2}{f} df - \int_{-\infty}^0 \frac{|\Psi(p-f)|^2}{f} df < \infty, \quad (1)$$

then for any signal  $x \in L^2(\mathcal{R})$ ,

$$\frac{1}{C_\psi^p} \int \int_{-\infty}^\infty \left| \text{GST}_x^{(p,1)}(t, f) \right|^2 \frac{dt df}{|f|} = \|x\|_2^2.$$

- The representation  $\frac{1}{C_\psi^p} \left| \text{GST}_x^{(p,1)}(t, f) \right|^2$  offers an energy distribution in the time-frequency domain with the measure  $\frac{dt df}{|f|}$ .
- If  $q \neq 1$ , the GST with an arbitrary window function generally does not satisfy the resolution of the identity formula.

# A practical example of the GST-based energy distribution

## Theorem: A Modified Gaussian Window

For

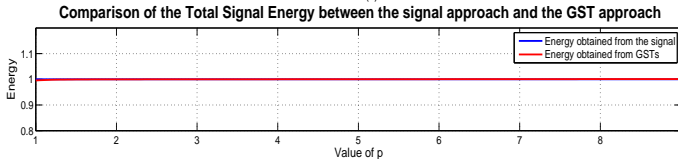
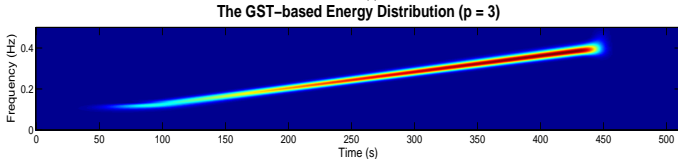
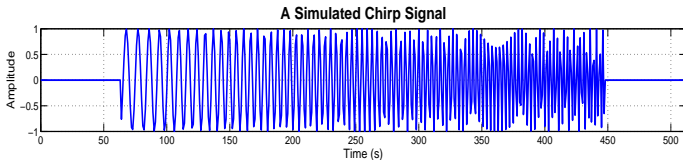
$$\Psi_g^p(f) = \frac{1}{K^p} \left( (1 + e^{-8\pi^2 p^2}) W_g(f) - e^{-2\pi^2 p^2} W_g(f + p) - e^{-2\pi^2 p^2} W_g(f - p) \right)$$

where  $W_g(f) = e^{-2\pi^2 f^2}$  and  $K^p = (1 - e^{-4\pi^2 p^2})^2$ , we then have

$$C_\Psi^p = \int_0^\infty \frac{|\Psi(f - p)|^2}{f} df - \int_{-\infty}^0 \frac{|\Psi(p - f)|^2}{f} df < \infty.$$

- The GST with the modified Gaussian window function provides a valid energy distribution.

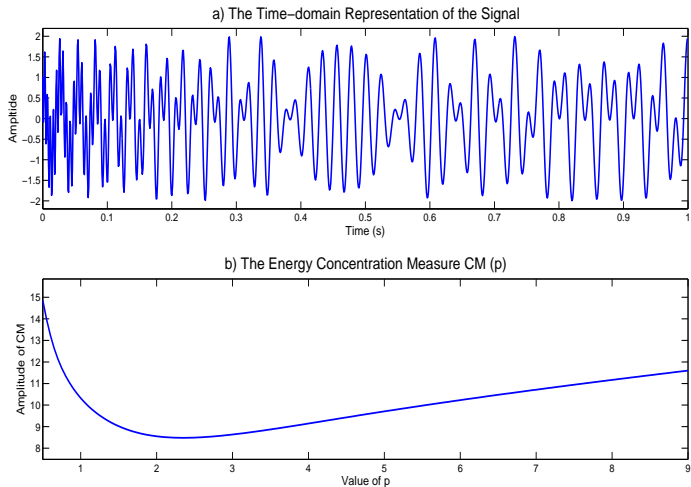
# An illustration of the GST-based energy distribution



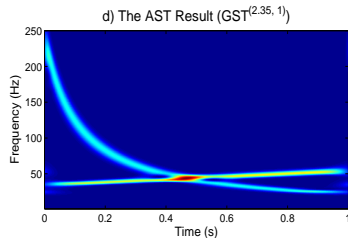
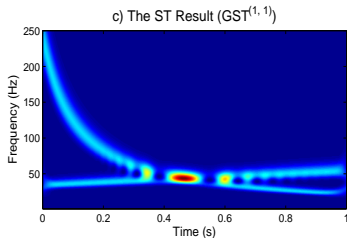
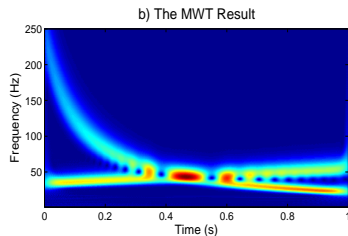
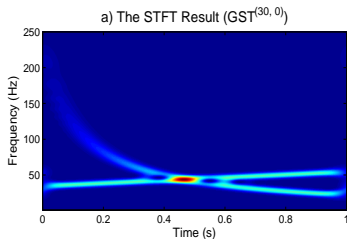
# Procedure to determine the adaptive Stockwell transform

- Step 1.** For each  $p$  from a given set, calculate the  $C_{\psi}^p$  and the GST  $GST_x^{(p,1)}(t, f)$  of the signal  $x$ .
- Step 2.** Compute the concentration measure
- $$CM_2(GST(p)) = \int \int \frac{|GST_x^{(p,1)}(t, f)|}{\sqrt{C_{\psi}^p |f|}} dt df.$$
- Step 3.** Determine the optimal value of  $p$  through minimizing the measure  $CM_2(GST(p))$ .
- Step 4.** The adaptive Stockwell transform associated with the signal  $x$  is given by  $AST_x(t, f) = GST_x^{(p_{opt}, 1)}(t, f)$ .

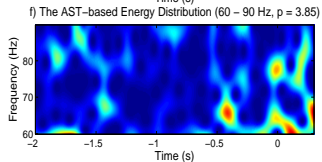
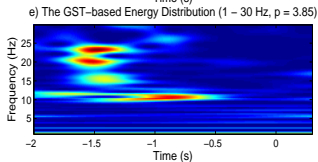
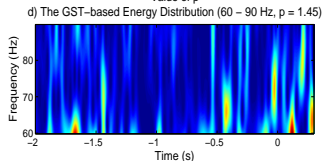
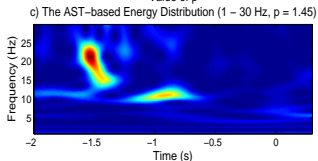
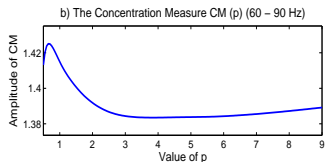
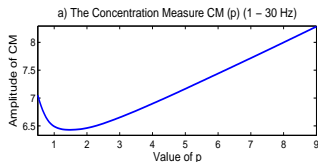
# A simulation example



# A simulation example



# A MEG example





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# GST-based measures

- 1 The GST-based power spectrum: reveals the time-varying spectral characteristics of a single time series.
- 2 The GST-based coherence function: describes the dynamics of linear interrelations between a pair of time series.
- 3 The GST-based phase-locking statistic: measures the phase synchronization between a pair of time series.

# The evolutionary spectrum (ES)

- Developed by M.B. Priestley, which presents a theoretical time-frequency domain measure of locally stationary time series.
- An approximation form

$$ES_{xx}(t, f) = \mathcal{F}_{\tau \rightarrow f} \{ \Gamma_{xx}(t, \tau) \otimes_t g(t, \tau) \},$$

where  $\Gamma_{xx}(t, \tau)$  is the autocorrelation function given by

$$\Gamma_{xx}(t, \tau) = E \left\{ \left( x^* \left( t - \frac{1}{2}\tau \right) x \left( t + \frac{1}{2}\tau \right) \right) \right\}.$$

and  $g(t, \tau)$  is a two-dimensional localization function.

# The power spectrum based on the Cohen's class distributions (CCDs)

- The Cohen's class distribution function was introduced by L. Cohen (1966), which is able to represent all bilinear TFRs.
- The CCD-based power spectrum:

$$\begin{aligned} TS_{xx}^{(\text{Cohen})}(t, f) &= E\{\text{CCD}_x(t, f)\} \\ &= \mathcal{F}_{\tau \rightarrow f}\{\Gamma_{xx}(t, \tau) \otimes_t \Phi(t, \tau)\}. \end{aligned}$$

where  $\Phi(t, \tau)$  is the time-lag kernel function.

- $\Phi(t, \tau)$  performs as the localized window function, which implies that any CCD-based power spectrum can be interpreted as an estimated evolutionary spectrum.

# The time-varying power spectrum based on the GST

## Definition: The GST-based Power Spectrum

$$TS_{xx}^{(GST)}(t, f) = E \left\{ GST_x^{(p,q)}(t, f) \cdot \overline{GST_x^{(p,q)}(t, f)} \right\}.$$

- The GST-based power spectrum presents a time-frequency domain measure of time series power.
- By choosing  $q = 1$  and  $p = p_{opt}$ , this definition leads to the AST-based power spectrum.

# The interpretation of the GST-based power spectrum

## Theorem: The Extended Cohen's Class of the GST

$$TS_{xx}^{(GST)}(t, f) = \mathcal{F}_{\tau \rightarrow f} \left\{ \Gamma_{xx}(t, \tau) \otimes_t \tilde{\Phi}^{(p,q)}(t, \tau; f) \right\}$$

with the frequency-dependent time-lag kernel function

$$\tilde{\Phi}^{(p,q)}(t, \tau; f) = \frac{f^{2q}}{p^2} \psi \left( \frac{|f|^q (-t - \frac{1}{2}\tau)}{p} \right) \overline{\psi \left( \frac{|f|^q (-t + \frac{1}{2}\tau)}{p} \right)}.$$

- The GST-based power spectrum presents an advanced estimation of the evolutionary spectrum with a frequency-dependent localization function.

# The coherence function

- A frequency domain measure of the interrelation between a pair of stationary time series.
- The cross-correlation function:

$$\Gamma_{xy}(t, \tau) = E \left\{ \left( y^* \left( t - \frac{1}{2}\tau \right) x \left( t + \frac{1}{2}\tau \right) \right) \right\}.$$

- The coherence function:

$$C_{xy}(f) = \frac{|S_{xy}(f)|^2}{S_{xx}(f) \cdot S_{yy}(f)}, \quad \text{if } S_{xx}(f) \cdot S_{yy}(f) \neq 0.$$

where  $S_{xy}(f) = \mathcal{F}_{\tau \rightarrow f} \{ \Gamma_{xy}(\tau) \}.$

# The GST-based coherence function

## Definition: The GST-based Coherence

For two time series  $x(t)$  and  $y(t)$ , the time-varying coherence based on the generalized Stockwell transform is defined as

$$TC_{xy}^{(GST)}(t, f) = \frac{|TS_{xy}^{(GST)}(t, f)|^2}{TS_{xx}^{(GST)}(t, f) \cdot TS_{yy}^{(GST)}(t, f)},$$

if  $TS_{xx}^{(GST)}(t, f) \cdot TS_{yy}^{(GST)}(t, f) \neq 0$ .

Note that  $0 \leq TC_{xy}^{(GST)}(t, f) \leq 1$ .

- The GST-based coherence provides a time-frequency domain measure of the interrelation characteristic, which can be interpreted as an advanced estimation of the evolutionary coherence.



# Phase synchronization

- The coherence function measures the linear interrelation between two time series, *i.e.*,

$$y(t) = \int_{-\infty}^{\infty} H(\tau)x(t - \tau)d\tau,$$

which does not separate the effects of amplitude and phase in the interaction.

- The phase synchronization evaluates the nonlinear interactions based on the synchronization of the instantaneous phase, *i.e.*,

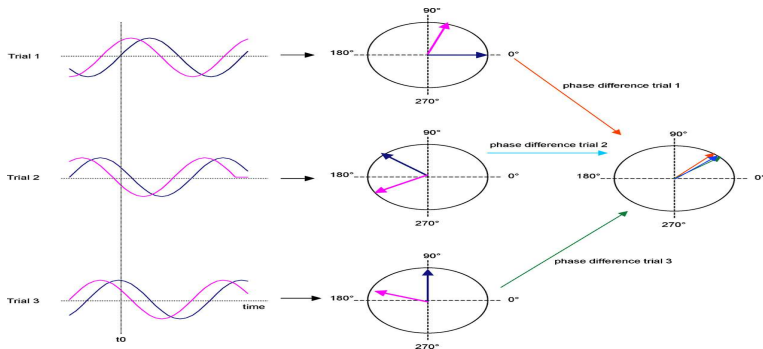
$$E\{\phi_x(t) - \phi_y(t)\} = \text{const.}$$

# The phase-locking statistic (PLS)

- Proposed by Lachaux, et al. (1999) as

$$\text{PLS}_{xy}(t) = \left| E\{e^{i(\phi_x(t) - \phi_y(t))}\} \right|,$$

where  $\phi_x(t)$  and  $\phi_y(t)$  are instantaneous phases of the analytic signals.



# The instantaneous phase of the GST

## Theorem: An Explicit Approximation Formula of the GST

Let the signal  $x(t) = a(t)e^{j\phi(t)}$ , with  $\psi(t) = \frac{1}{\sqrt{2\pi}}e^{-\frac{t^2}{2}}$ , we have

$$\text{GST}_x^{(p,q)}(t, f) \approx A^{(\text{GST})}(t, f)e^{j\phi^{(\text{GST})}(t, f)},$$

where

$$\phi^{(\text{GST})}(t, f) \approx \phi(t) - j2\pi tf.$$

- The GST holds the absolutely referenced phase information.
- An approximation formula of the ST with a general window function can be found in (Guo, Molahajloo and Wong, 2010).

# The GST-based phase-locking statistic (PLS)

## Definition: The GST-based Phase-locking Statistic

For two time series  $x(t)$  and  $y(t)$ , the phase-locking statistic based on the generalized Stockwell transform is defined as

$$\text{PLS}_{xy}^{(\text{GST})}(t, f) = \left| \mathbb{E} \left\{ \frac{\text{GST}_x^{(p,q)}(t, f) \cdot \overline{\text{GST}_y^{(p,q)}(t, f)}}{|\text{GST}_x^{(p,q)}(t, f)| \cdot |\text{GST}_y^{(p,q)}(t, f)|} \right\} \right|$$



$$\text{PLS}_{xy}^{(\text{GST})}(t, f) \approx \left| \mathbb{E} \left\{ e^{i[\phi_x(t) - \phi_y(t)]} \right\} \right|.$$

Phases  $\phi_x(t)$  and  $\phi_y(t)$  here are the phases of the filtered signals around the considered frequency  $f$ .

- The GST-based PLS provides the phase synchronization measure across a wide range of frequencies.

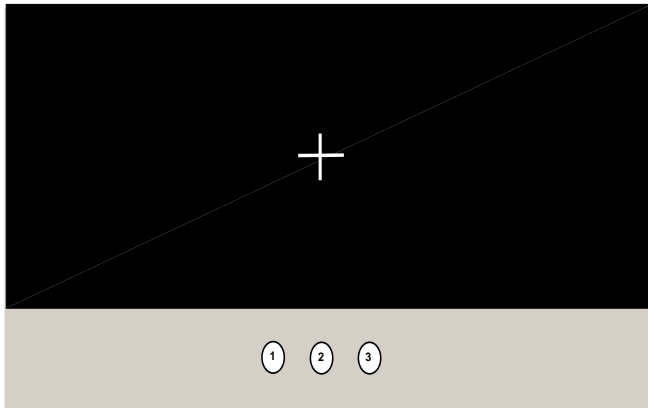
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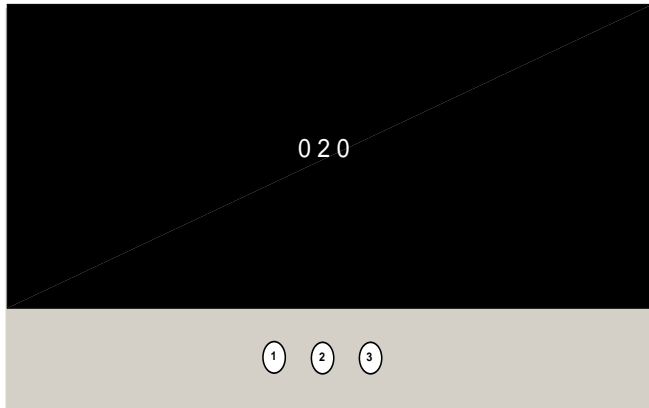
# The Multi-source Interference Task (MSIT)

- Designed to study normal human cognition and psychiatric pathophysiology (Bush, et al., 2003).
- A behavioral experiment involving tasks at multiple levels of difficulty.
- There are two types of task trials in the MSIT: control trials and interference trials.

# The control trials of the MSIT

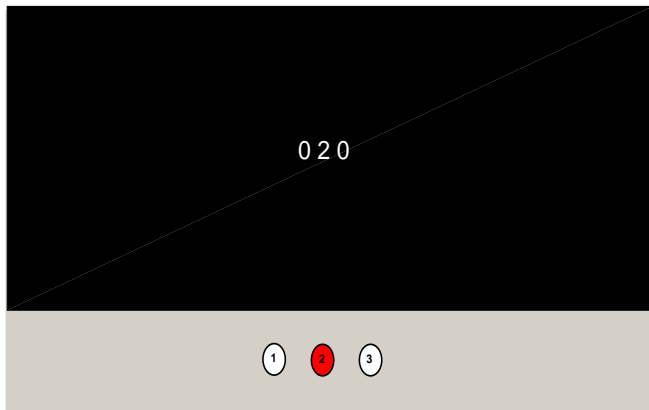


# The control trials of the MSIT

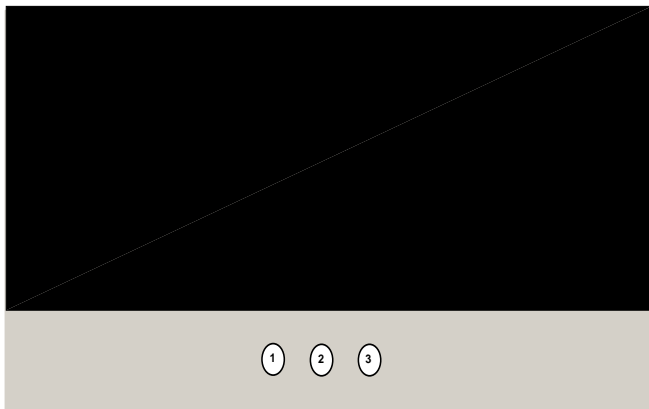




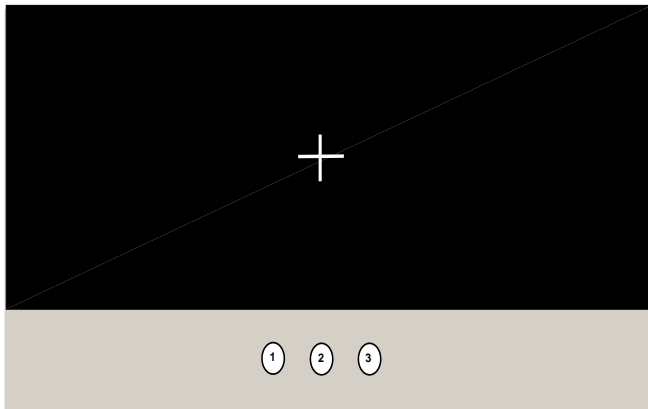
# The control trials of the MSIT



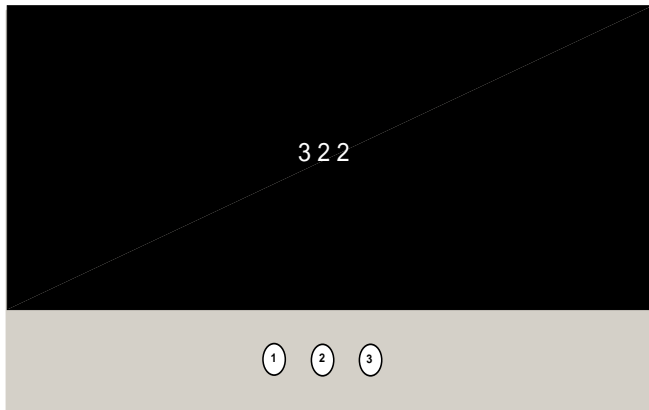
# The control trials of the MSIT



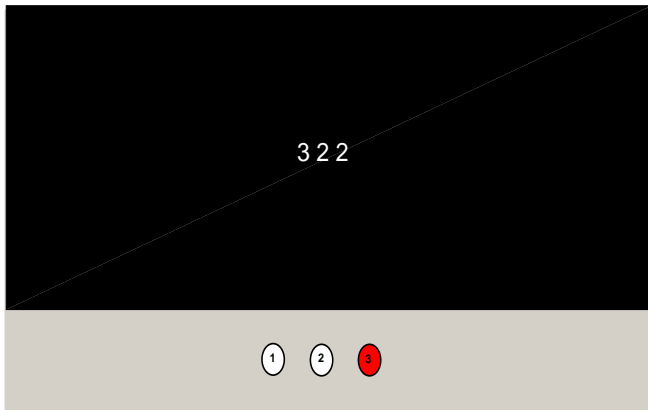
# The interference trials of the MSIT



# The interference trials of the MSIT



# The interference trials of the MSIT



# Acquisition and Preprocessing

- The right-handed subjects (6 males, 4 females; mean age: 31) performed the MSIT.
- The brain activity was measured using a 275 channel whole-head MEG.
- Around 80 control and 80 interference trials were recorded continuously for each subject.
- Data were epoched at -3.5s to 0.5s with respect to the response at time = 0s.
- Trials with incorrect response were removed before subsequent analysis.

# Functional activities of motor cortices

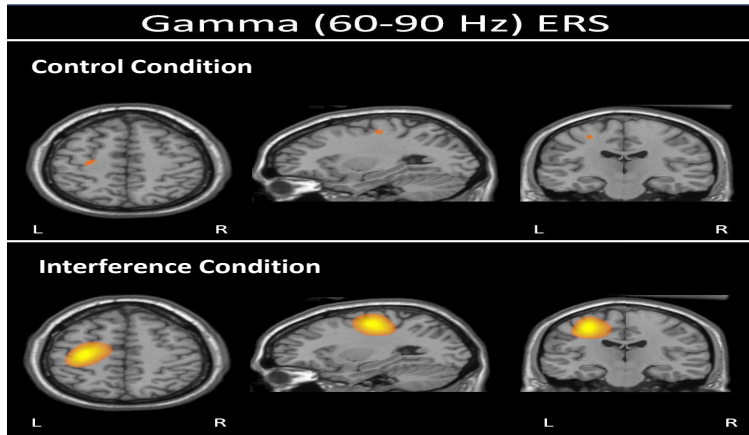
- 1 Study the influence of the task difficulty on functional activities at the contralateral motor cortex (Mlc) in the gamma frequency band ([60 - 90 Hz]).
  - 2 Study functional couplings between the contralateral motor cortex and the ipsilateral motor cortex (Mli) in the alpha ([8 - 12 Hz]) and beta ([12 - 30 Hz]) frequency bands.
- **To improve the accuracy, we apply the proposed adaptive measures to the studies. Note that the resolution of these measures automatically adjusts to the specific frequency bands of interests.**

# Studying gamma-band activities at the Mlc

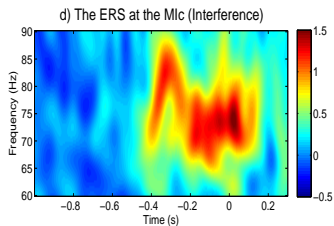
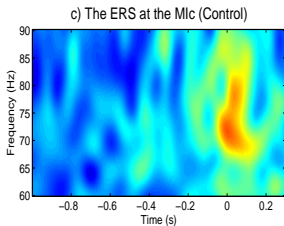
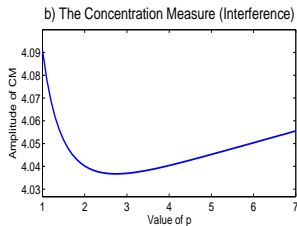
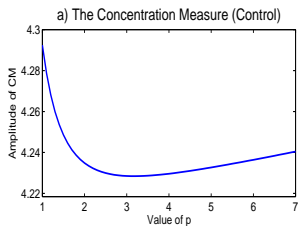
- A differential minimum-variance **beamformer algorithm (SAM)** (60-90 Hz) was applied for source localization at the Mlc.
- The motor activity at the gamma-frequency band (**event-related synchronization (ERS)**) was calculated based on the AST-based power spectrum for any group of MEG trials.
- The **t-statistic** was computed to assess the difference of gamma-band Mlc activities in time between two groups of trials.
- The statistical significance was further examined using a **permutation test** ( $N = 2000$ ,  $\alpha = 0.05$ ) and corrected following the single threshold test.



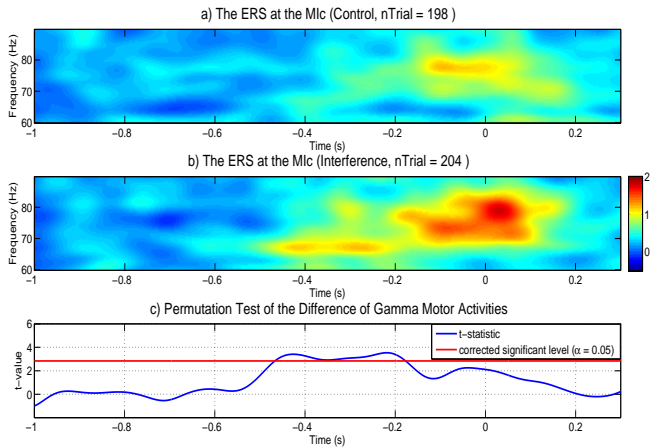
# The SAM result



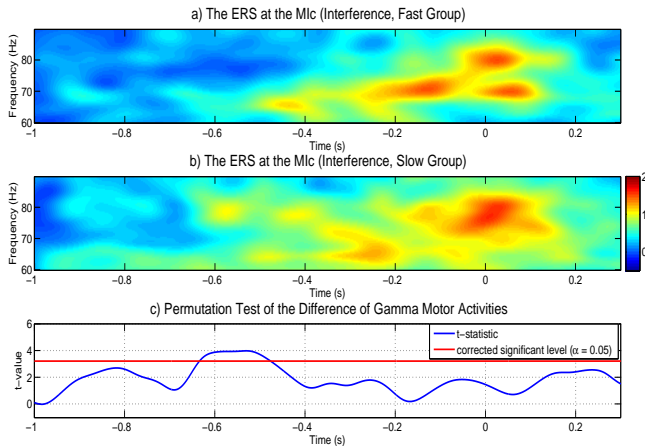
# The TF results of a single subject



# Group analysis: control trials VS interference trials



# Group analysis: fast trials VS slow trials (interference)



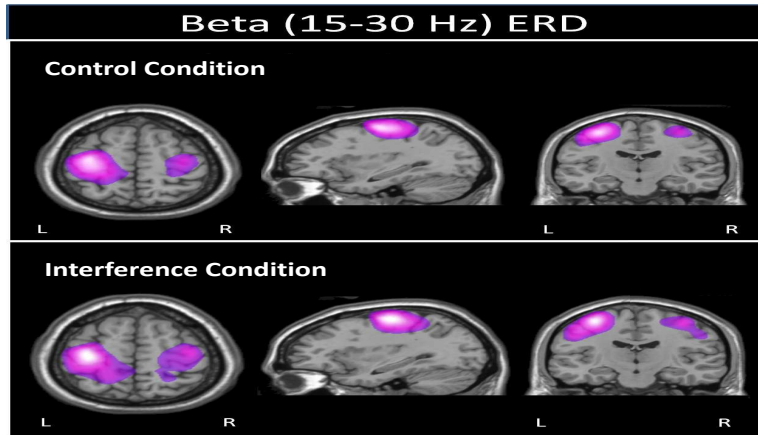
# Our findings

- More gamma-band activities at the Mlc were observed for interference trials compared to control trials,
- More gamma-band activities at the Mlc were observed for slow trials compared to fast trials under the interference condition.
- Both the task condition and the RT information provide measures for the task difficulty in the MSIT. Therefore, our results suggest that more gamma-band Mlc activities are needed for dealing with more difficult tasks.

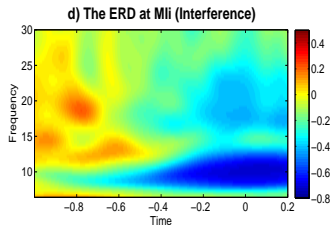
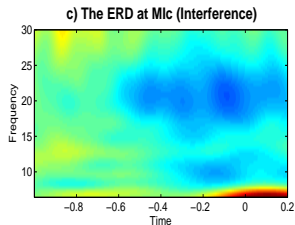
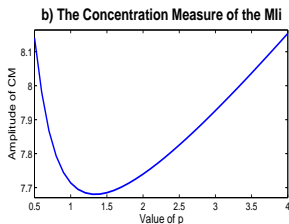
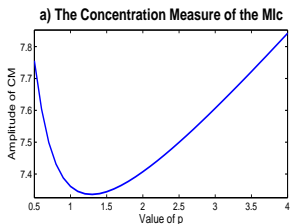
# Study functional couplings between the Mlc and the Mli

- A differential SAM (15 - 30 Hz) was applied for source localization around the Mlc and the ipsilateral motor cortex (Mli).
- The motor activity at frequency bands of alpha and beta (event-related desynchronization (ERD)) was calculated based on the AST-based power spectrum for each group of MEG trials.
- The AST-based coherence and PLS were computed to measure the interactions between the Mlc and the Mli in the time-frequency domain.
- The statistical significance was further examined using a bootstrap test ( $N = 500, \alpha = 0.05$ ).

# The SAM result



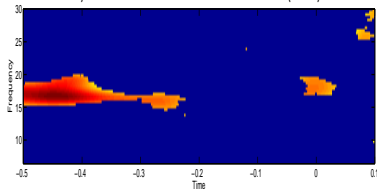
# The TF results of a single subject (interference)



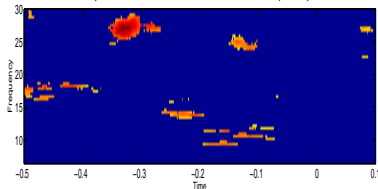


# Grand averaged results: the functional coupling

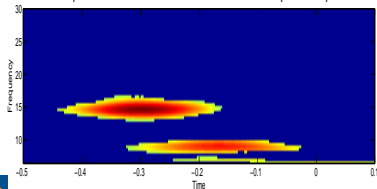
a) The AST-based Coherence between the Mic and the Mli (Control)



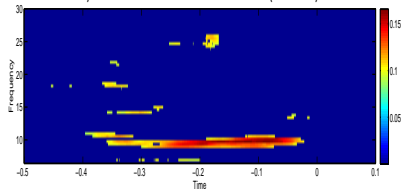
a) The AST-based PLS between the Mic and the Mli (Control)



b) The AST-based Coherence between the Mic and the Mli (Interference)



b) The AST-based PLS between the Mic and the Mli (Interference)



# Our findings

- Both of the AST-based coherence and PLS results shown functional couplings between the Mlc and the Mli, which were predominant for interference trials rather than control trials.
- Our results suggest that the functional connectivity between the Mlc and the Mli at low frequencies ([8 - 15 Hz]) mainly appears when subjects perform interference trials under the MSIT.

# Outline

- 1 Introduction
- 2 An Adaptive Multi-resolution Time-frequency Representation
- 3 Time-varying Spectral Measures for Analyzing Brain Time Series
- 4 Studying Motor Activities using the Multi-source Interference Task
- 5 Summary and Future Work

# Summary

- Motivation: investigate functional activities of motor cortices using the MSIT

**Multi-Source Interference Task (MSIT)**

**Control trial example**

0.5 s

+

Task: "Which one of these numbers is not like the others?"

In this example, the 2 is different than the 0s, so push button 2. Note that for control trials, the targets *always* differ in magnitude, and therefore they are the easier targets. Thus, response congruency stimuli are relatively easy to perform.

3 s

020

Interact response

0.5 s

Total set of possible control stimuli: (000, 001, 002).

**Interference trial example**

0.5 s

+

Task: "Which one of these numbers is not like the others?"

In this example, the 3 is different than the 0s, so push button 3. Note that for interference trials, the targets *never* match the response location, and the harder stimuli are *always* potential targets. Thus, response incongruent stimuli are relatively difficult to perform.

3 s

322

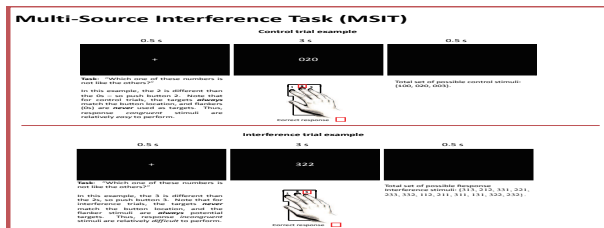
Interact response

0.5 s

Total set of possible Response Interference stimuli: (313, 312, 331, 331, 233, 332, 112, 211, 311, 131, 322, 232).

# Summary

- Motivation: investigate functional activities of motor cortices using the MSIT

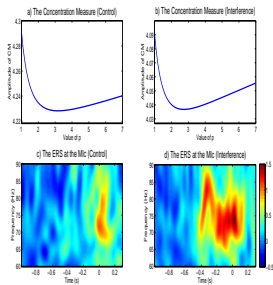


- A new time-frequency analysis tool

$$GST_{\psi}^{(p,q)}(t, f) \longrightarrow GST_{mGaussian}^{(p,1)}(t, f) \longrightarrow AST_{mGaussian}^{(p_{opt},1)}(t, f)$$

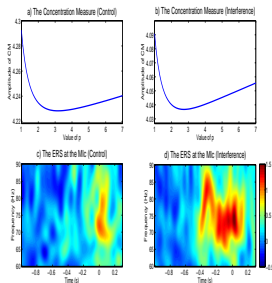
# Summary

- The AST-based **power spectrum**:  
a time-varying power spectrum

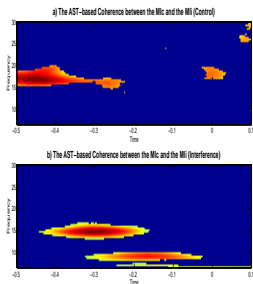


# Summary

- The AST-based **power spectrum**:  
a time-varying power spectrum

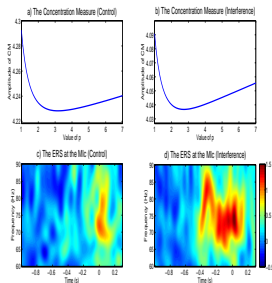


- The AST-based **coherence**:  
a linear interrelation measure

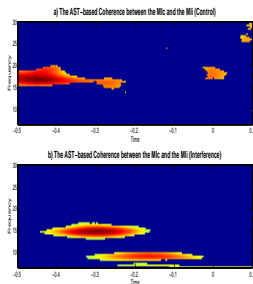


# Summary

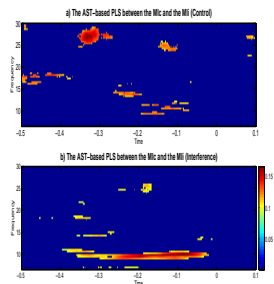
- The AST-based **power spectrum**:  
a time-varying power spectrum



- The AST-based **coherence**:  
a linear interrelation measure



- The AST-based **PLS**:  
a phase synchronization measure



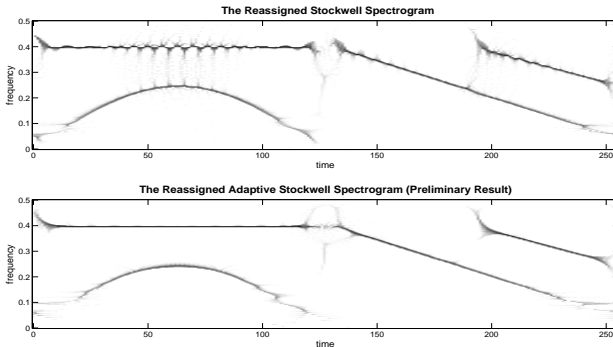


# Future work

- 1 Find energy distribution forms of the generalized Stockwell transforms with arbitrary values of  $q$ .
- 2 Extend the adaptive approach to multi-dimensional Stockwell transforms.
- 3 Develop advanced time-frequency domain measures for studying cross-frequency coupling and unidirectional interactions between brain areas.

# Future work

- Extract instantaneous frequencies of non-stationary signals



# Thank you

- I would like to thank **Dr. William Gaetz** from the Children's Hospital of Philadelphia for providing the experimental data and **Professor Hongmei Zhu** from York University for guiding this research work.

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- I would like to thank **Dr. William Gaetz** from the Children's Hospital of Philadelphia for providing the experimental data and **Professor Hongmei Zhu** from York University for guiding this research work.
- **Question ?**



# Reference I



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