

Recent Advances in Biomagnetic Signal Processing and Source Localization  
Satellite Symposium for Biomag 2006  
Vancouver, August, 20, 2006

# **Prewhitening Signal Covariance Estimation and Prewhitening Beamforming**

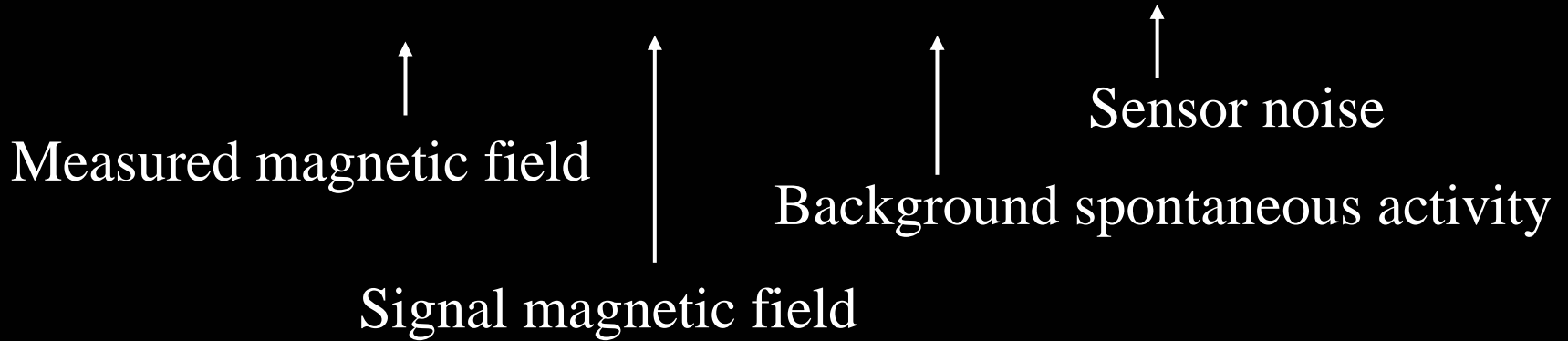
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# Data model

$$\mathbf{b}(t) = \mathbf{b}_s(t) + \mathbf{b}_I(t) + \mathbf{n}(t)$$



## Definitions

Data covariance:  $\mathbf{R} = \langle \mathbf{b}(t) \mathbf{b}^T(t) \rangle$

Signal covariance:  $\mathbf{R}_S = \langle \mathbf{b}_s(t) \mathbf{b}_s^T(t) \rangle$

Interference plus noise covariance:  $\mathbf{R}_{i+n} = \langle (\mathbf{b}_I(t) + \mathbf{n}(t)) (\mathbf{b}_I(t) + \mathbf{n}(t))^T \rangle$

# Two-condition experiments (ideal scenario)

Task:  $\mathbf{b}(t) = \mathbf{b}_s(t) + \mathbf{b}_I(t) + \mathbf{n}(t)$

Control:  $\mathbf{b}_C(t) = \mathbf{b}_I(t) + \mathbf{n}(t)$



Covariance matrix relations

Task:  $\mathbf{R} = \mathbf{R}_S + \mathbf{R}_{i+n}$

Control:  $\mathbf{R}_C = \mathbf{R}_{i+n}$

Problem:

How to obtain interference-free source reconstruction

How to estimate  $\mathbf{R}_S$  using  $\mathbf{R}$  and  $\mathbf{R}_C$

## Existing approach: image-based subtraction (with $t$ test)

Define  $s(\mathbf{r})$  : source reconstruction from  $\mathbf{b}(t)$

$s_c(\mathbf{r})$  : source reconstruction from  $\mathbf{b}_c(t)$

Calculate  $\Delta s(\mathbf{r}) = s(\mathbf{r}) - s_c(\mathbf{r})$

**This approach works well when SIR is high but becomes less effective when large interference exists.**

## Naive approach: covariance-based subtraction

Estimate  $\mathbf{R}_S$  using  $\hat{\mathbf{R}}_S = \mathbf{R} - \mathbf{R}_C$

**This approach has many problems, and generally does not work well**

# Prew whitening estimation of signal covariance

Calculate  $\tilde{\mathbf{R}} = \mathbf{R}_C^{-1/2} \mathbf{R} \mathbf{R}_C^{-1/2}$

(Tilde is used to indicate the prewhitened version of a matrix)

$$\tilde{\mathbf{R}}_S = \sum_{j=1}^Q \gamma_j \mathbf{u}_j \mathbf{u}_j^T \Rightarrow \tilde{\mathbf{R}} = \underbrace{\sum_{j=1}^Q (\gamma_j + 1) \mathbf{u}_j \mathbf{u}_j^T}_{\uparrow\uparrow} + \sum_{j=Q+1}^M \mathbf{u}_j \mathbf{u}_j^T.$$

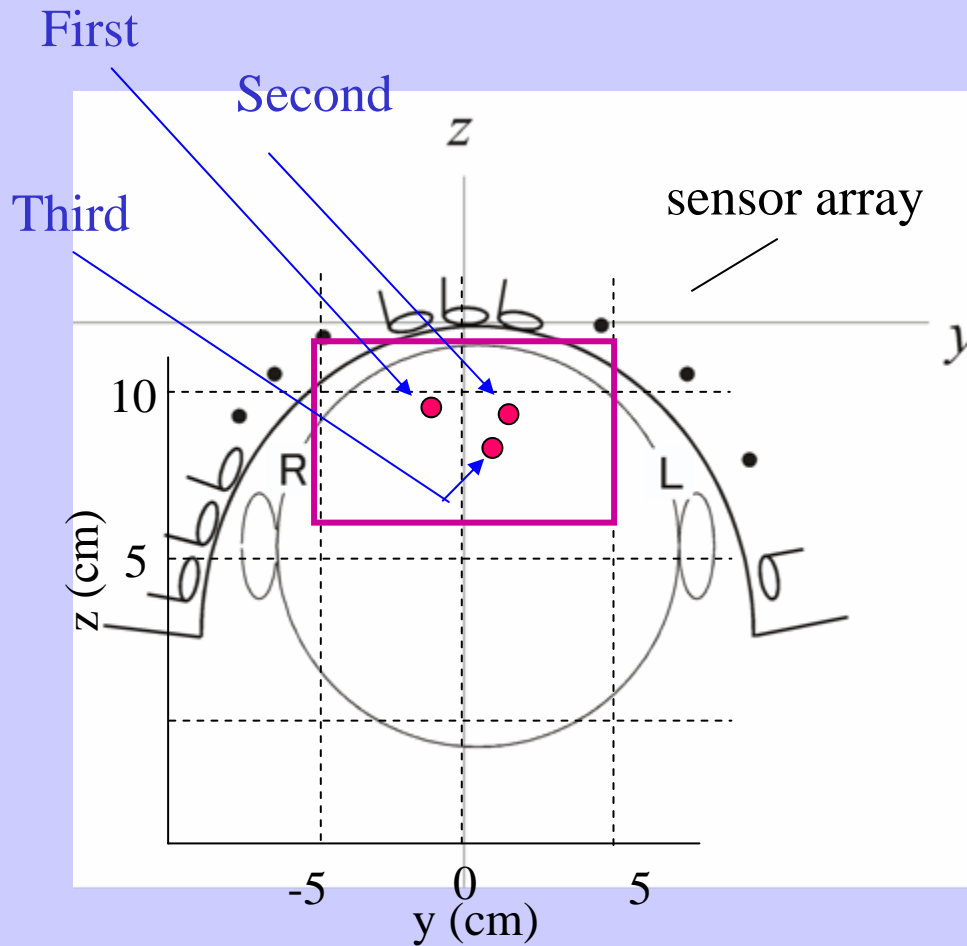
$\tilde{\mathbf{R}}$  has signal-level eigenvalues greater than 1,  
and their eigenvectors are equal to those of  $\tilde{\mathbf{R}}_S$

Signal covariance estimation

$$\hat{\mathbf{R}}_S = \mathbf{R}_C^{1/2} \left[ \mathbf{U}_S \mathbf{U}_S^T (\tilde{\mathbf{R}} - \mathbf{I}) \right] \mathbf{R}_C^{1/2} = \mathbf{R}_C^{1/2} \left[ \sum_{j=1}^Q (\gamma_j - 1) \mathbf{u}_j \right] \mathbf{R}_C^{1/2}$$

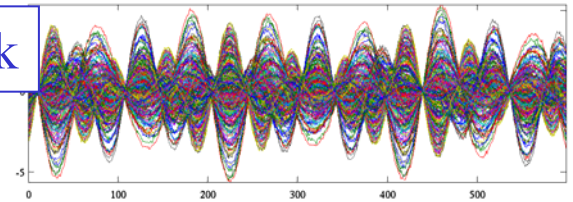
$\mathbf{U}_S \uparrow$   
 $\mathbf{U}_S = [\mathbf{u}_1, \dots, \mathbf{u}_Q]$

# Computer simulation on two-condition multi-epoch measurements

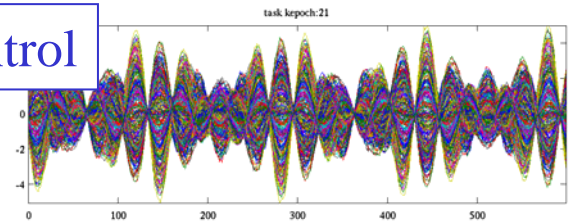


## Simulated signal recordings

Task

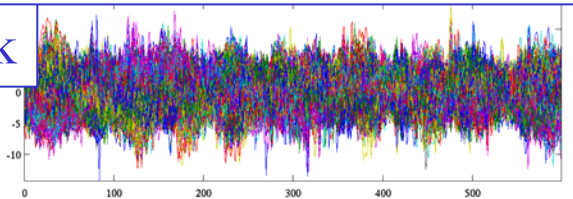


Control

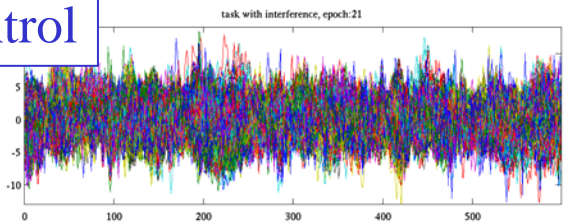


## Spontaneous MEG added

Task



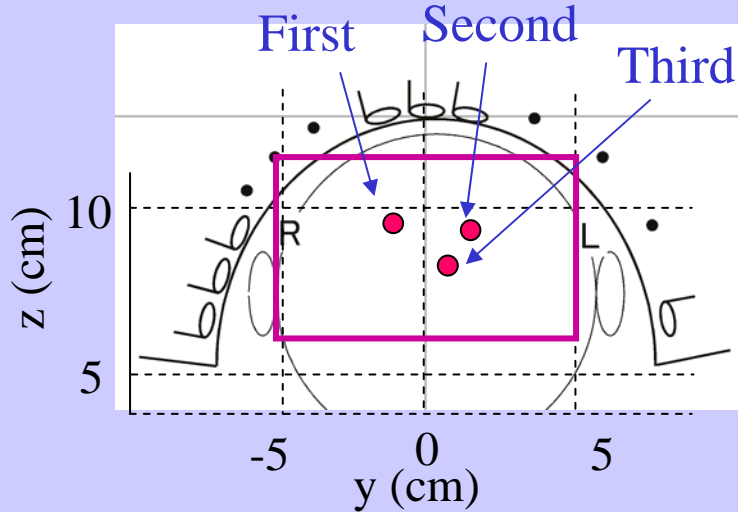
Control



A total 96 of these paired data sets were generated, and used for calculating  $R$  and  $R_C$ .

# Two-condition measurements-Computer Simulation

Signal-to-Interference Ratio: **0.5**



Relative source intensity

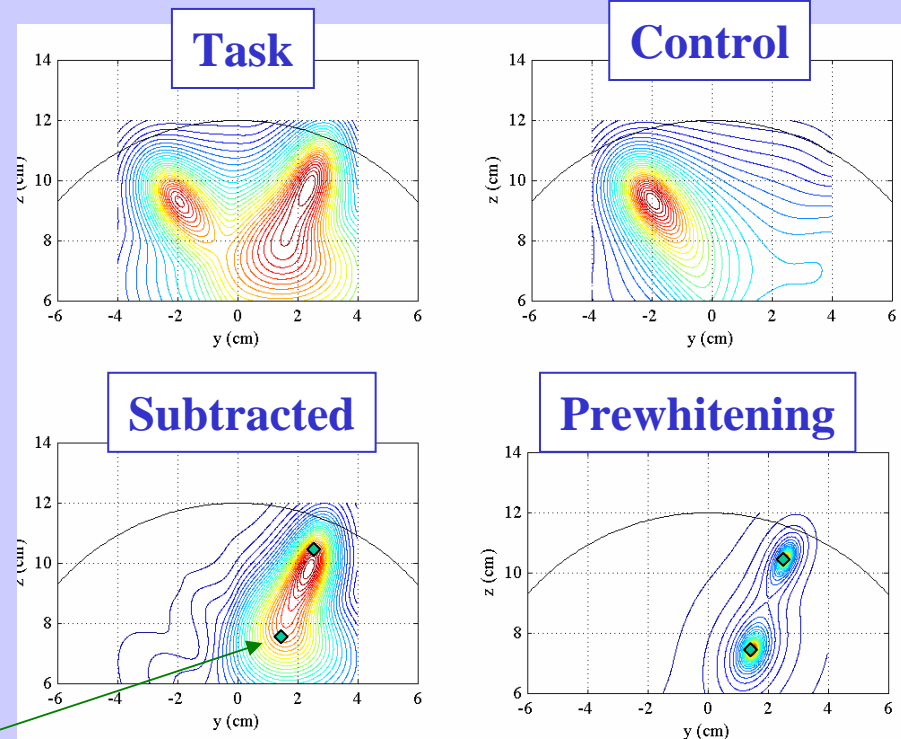
	control	task
First source	1	1
Second source	0	1
Third source	0	1

Minimum-variance source reconstruction

$$\hat{s}(\mathbf{r}, t) = \frac{1}{[\mathbf{l}^T(\mathbf{r})\mathbf{R}^{-1}\mathbf{l}(\mathbf{r})]}$$

$$\hat{s}_{PW}(\mathbf{r}, t) = \frac{1}{[\mathbf{l}^T(\mathbf{r})(\hat{\mathbf{R}}_S + \mu\mathbf{I})^{-1}\mathbf{l}(\mathbf{r})]}$$

source location



# Two-condition experiments: more realistic scenario

## - Scenario I

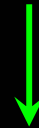
Some sources exist only in the control condition

Task:  $\mathbf{b}(t) = \mathbf{b}_s(t) + \mathbf{b}_I(t) + \mathbf{n}(t)$

Control:  $\mathbf{b}_c(t) = \mathbf{b}_I(t) + \mathbf{n}(t) + \mathbf{b}_\Delta(t)$



Signal from control-only sources



Task:  $\mathbf{R} = \mathbf{R}_S + \mathbf{R}_{i+n}$

Control:  $\mathbf{R}_C = \mathbf{R}_{i+n} + \mathbf{R}_\Delta$

covariance from control-only source:  $\mathbf{R}_\Delta = \langle \mathbf{b}_\Delta(t) \mathbf{b}_\Delta^T(t) \rangle$



# Scenario I-analysis

Define

$$\tilde{\mathbf{R}}_S = \sum_{j=1}^Q \gamma_j \mathbf{u}_j \mathbf{u}_j^T, \text{ and } \tilde{\mathbf{R}}_\Delta = \sum_{j=1}^P \beta_j \mathbf{v}_j \mathbf{v}_j^T$$

Then

$$\tilde{\mathbf{R}} = \tilde{\mathbf{R}}_S + \mathbf{I} - \tilde{\mathbf{R}}_\Delta = \sum_{j=1}^Q \gamma_j \mathbf{u}_j \mathbf{u}_j^T + \sum_{j=P+1}^M \beta_j \mathbf{v}_j \mathbf{v}_j^T + \sum_{j=1}^P (1 - \beta_j) \mathbf{v}_j \mathbf{v}_j^T$$

Thus

$$\begin{aligned} \tilde{\mathbf{R}} \mathbf{u}_j &= \left( \sum_{j=1}^Q \gamma_j \mathbf{u}_j \mathbf{u}_j^T \right) \mathbf{u}_j + \left( \sum_{j=P+1}^M \beta_j \mathbf{v}_j \mathbf{v}_j^T \right) \mathbf{u}_j + \left( \sum_{j=1}^P (1 - \beta_j) \mathbf{v}_j \mathbf{v}_j^T \right) \mathbf{u}_j \\ &= \gamma_j \mathbf{u}_j + \sum_{j=1}^P (1 - \beta_j) (\mathbf{v}_j^T \mathbf{u}_j) \mathbf{v}_j \end{aligned}$$

Prewhitening estimation  $\hat{\mathbf{R}}_S = \mathbf{R}_C^{1/2} \left[ \mathbf{U}_S \mathbf{U}_S^T (\tilde{\mathbf{R}} - \mathbf{I}) \right] \mathbf{R}_C^{1/2}$  is still effective if  $\text{span}(\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_Q) \perp \text{span}(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_P)$  approximately holds.

# Two-condition experiments: more realistic scenario

## - Scenario II

Target signal sources are active also in the control condition  
(Their intensities change between the two conditions)

Task:  $\mathbf{b}(t) = \mathbf{b}_s(t) + \mathbf{b}_I(t) + \mathbf{n}(t)$

Control:  $\mathbf{b}_c(t) = \mathbf{b}_s(t)' + \mathbf{b}_I(t) + \mathbf{n}(t)$



Task:  $\mathbf{R} = \mathbf{R}_S + \mathbf{R}_{i+n}$

Control:  $\mathbf{R}_C = \mathbf{R}_S' + \mathbf{R}_{i+n}$

$$\mathbf{R}_S' = \langle \mathbf{b}_s(t)' \mathbf{b}_s^T(t)' \rangle$$

## Scenario II-analysis

We finally have

$$\tilde{\mathbf{R}} = \tilde{\mathbf{D}}_P + \mathbf{I} - \tilde{\mathbf{D}}_N$$

where

$\mathbf{D}_P$  : covariance matrix from signal sources  
**stronger** in the task than in the control state.

$\mathbf{D}_N$  : covariance matrix from signal sources  
**weaker** in the task than in the control state.

The above covariance relationship is the same as that for Scenario I

We can still use  $\hat{\mathbf{D}}_P = \mathbf{R}_C^{1/2} \left[ \mathbf{U}_S \mathbf{U}_S^T (\tilde{\mathbf{R}} - \mathbf{I}) \right] \mathbf{R}_C^{1/2}$

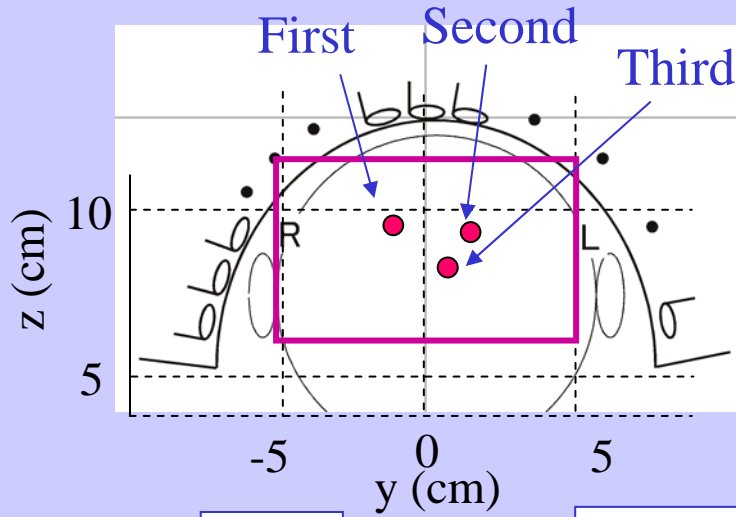
$\hat{\mathbf{D}}_N$  is obtained from flipped prewhitening where  $\tilde{\mathbf{R}}_C = \mathbf{R}^{-1/2} \mathbf{R}_C \mathbf{R}^{-1/2}$  is used.

# Scenario I and II--Numerical Experiments

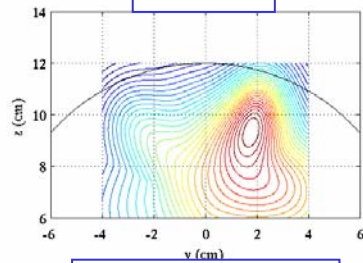
**Signal-to-Interference Ratio: 0.5**

**Relative source intensity**

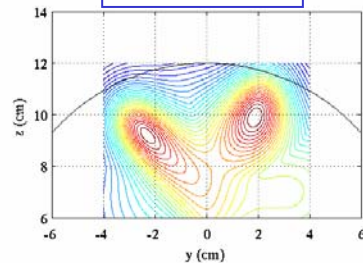
	control	task
First source	1	0.5
Second source	1	1
Third source	0.5	1



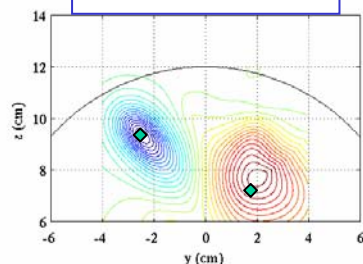
**Task**



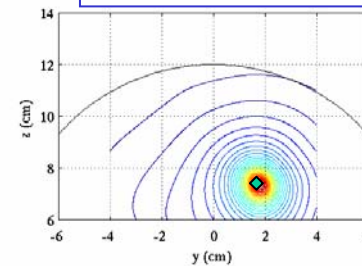
**Control**



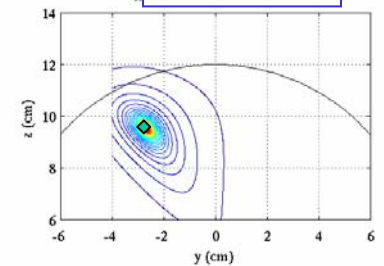
**Subtracted**



**Prewhitening**



**Flipped**

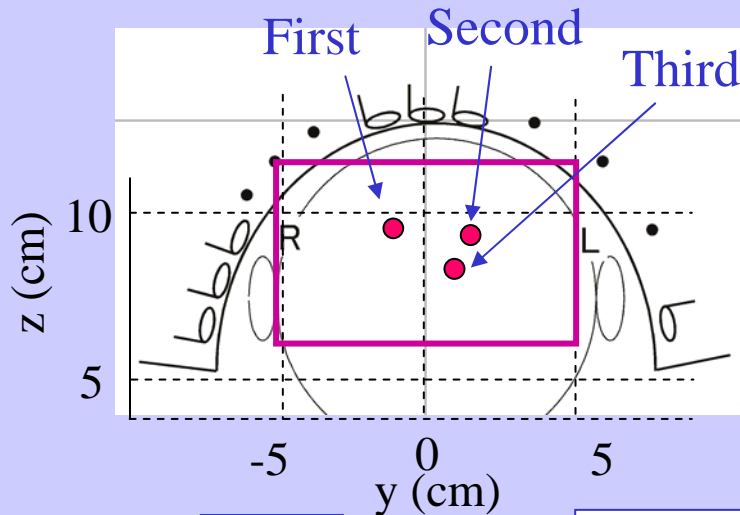


# Scenario I and II--Numerical Experiments

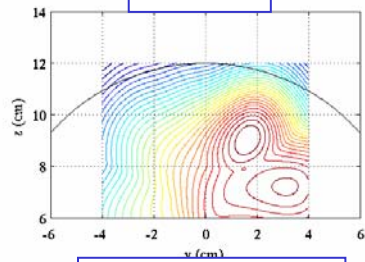
**Signal-to-Interference Ratio: 0.25**

**Relative source intensity**

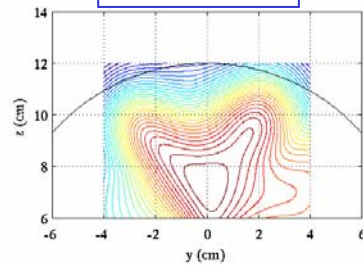
	control	task
First source	1	0.5
Second source	1	1
Third source	0.5	1



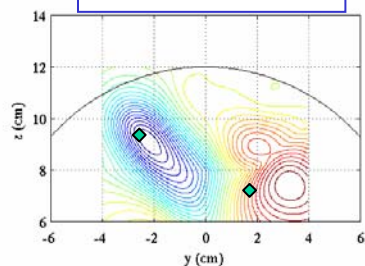
**Task**



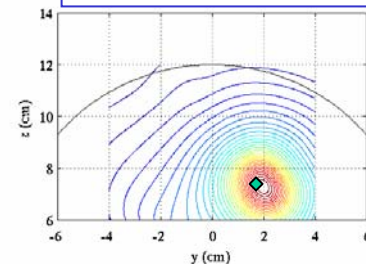
**Control**



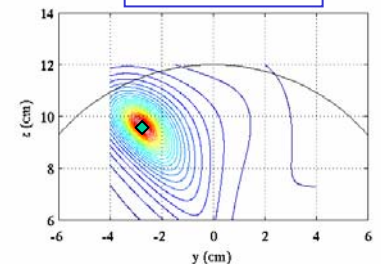
**Subtracted**



**Prewhitening**



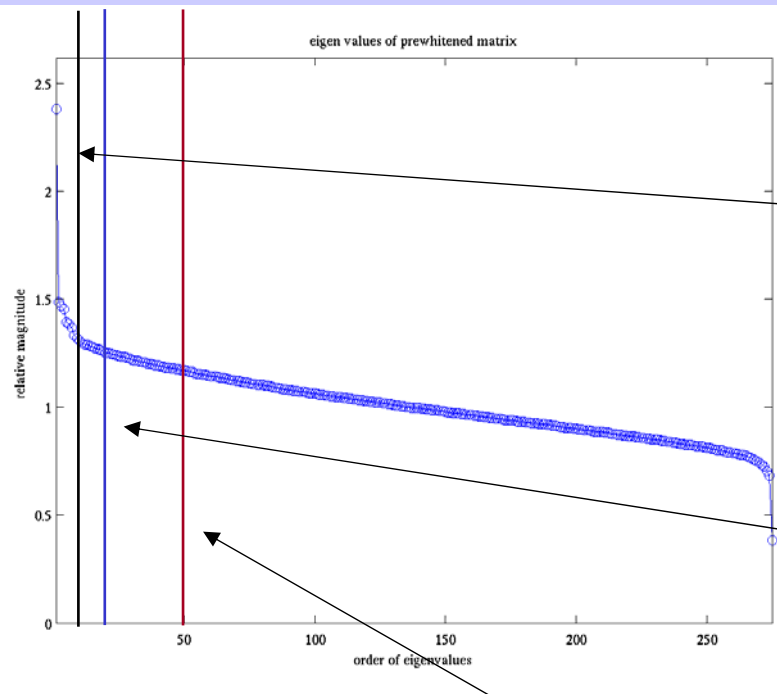
**Flipped**



# Robustness to overestimation of signal subspace dimension

Signal-to-Interference Ratio:0.5

Eigenvalue spectrum of  $\tilde{\mathbf{R}}$

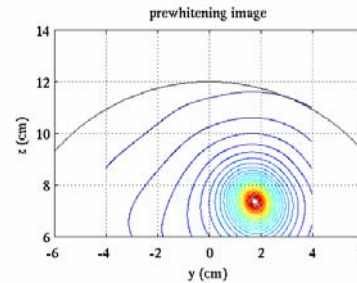


$Q = 10$

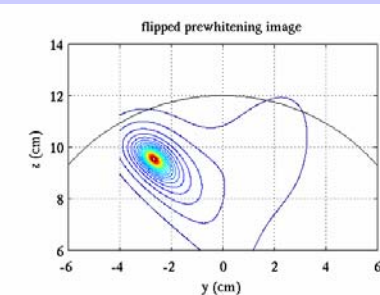
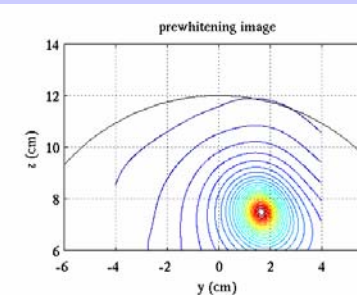
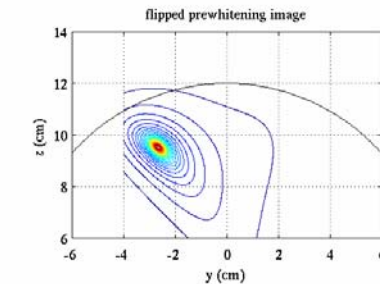
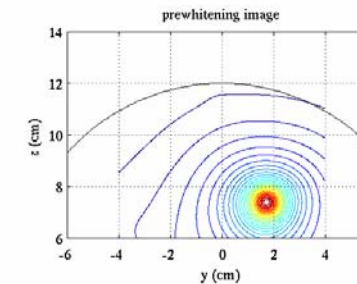
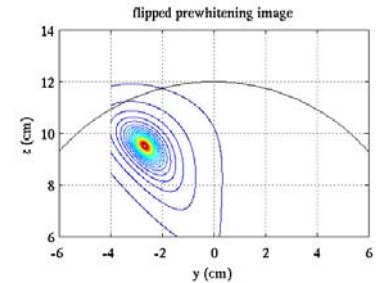
$Q = 20$

$Q = 50$

Prewhitening



Flipped



# Overestimation of signal-subspace dimension

$$\hat{\mathbf{R}}_S = \mathbf{R}_C^{1/2} \left[ (\mathbf{U}_S \mathbf{U}_S^T + \mathbf{U}_\varepsilon \mathbf{U}_\varepsilon^T) (\tilde{\mathbf{R}} - \mathbf{I}) \right] \mathbf{R}_C^{1/2} = \mathbf{R}_S + \Delta \mathbf{R}_S$$

Overestimated term

Error

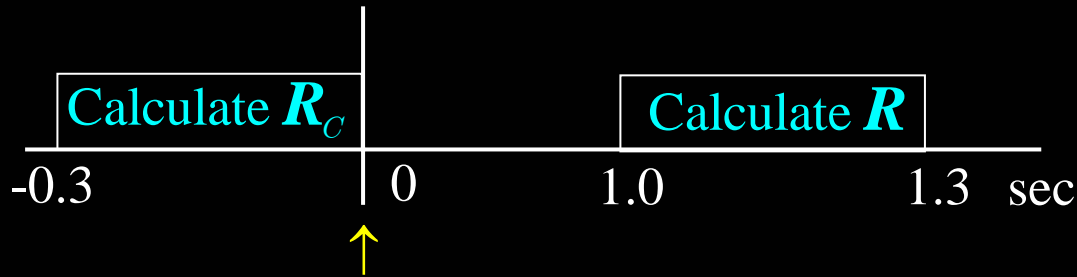
$$\Delta \mathbf{R}_S = \underbrace{\sum_{j=1}^Q \Delta \lambda_j \mathbf{e}_j \mathbf{e}_j^T}_{\text{signal subspace component}} + \underbrace{\sum_{j=Q+1}^M \Delta \mu_j \mathbf{e}_j \mathbf{e}_j^T}_{\text{noise subspace component}}$$

$$\hat{\mathbf{R}}_{S+n} = \mathbf{R}_S + \Delta \mathbf{R}_S + \mu \mathbf{I} = \sum_{j=1}^Q (\lambda_j + \Delta \lambda_j) \mathbf{e}_j \mathbf{e}_j^T + \sum_{j=Q+1}^M (\mu + \Delta \mu_j) \mathbf{e}_j \mathbf{e}_j^T$$

These terms modify source intensity

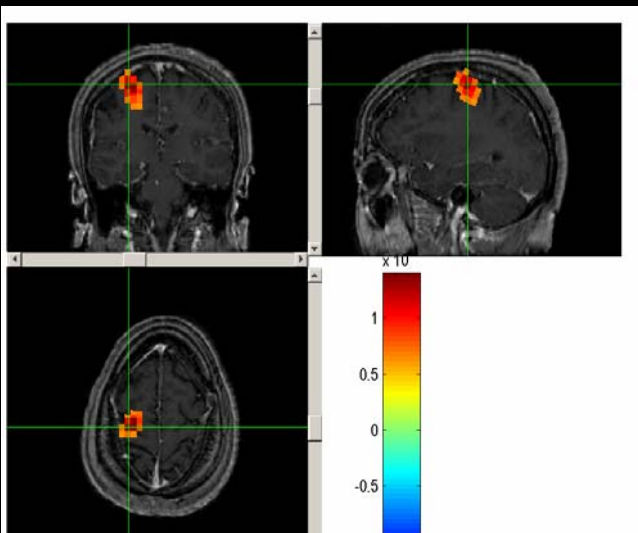
These terms modify regularization constant

# Hand-motor measurement

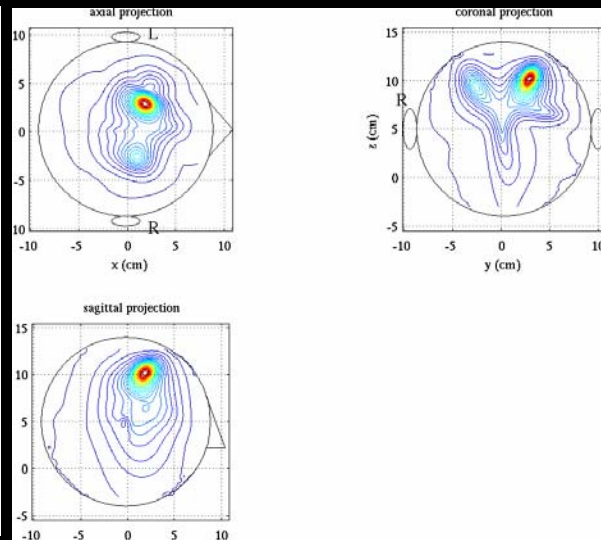


Voluntary right finger movement (every 5 sec)

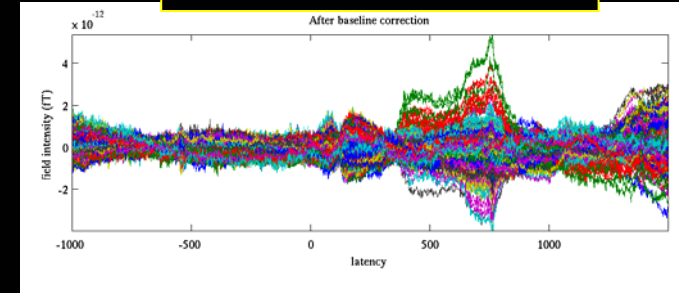
MRI overlay



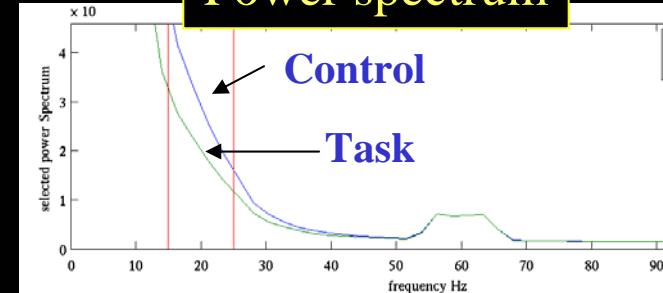
Prewhitening results



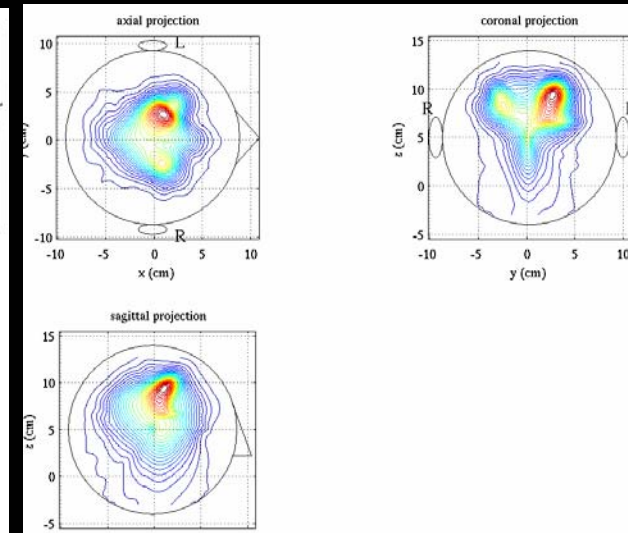
Typical raw epoch



Power spectrum

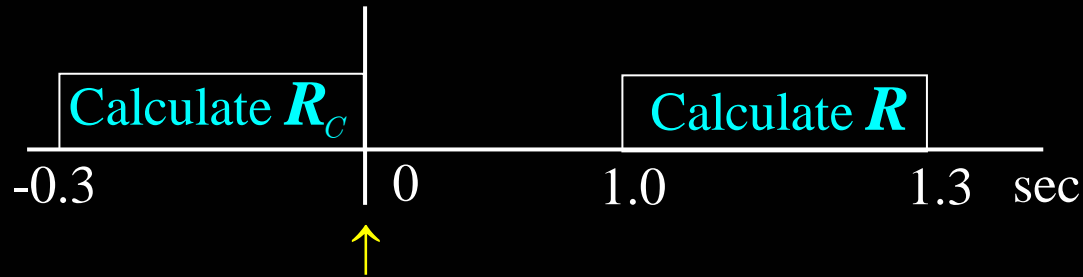


Subtraction results



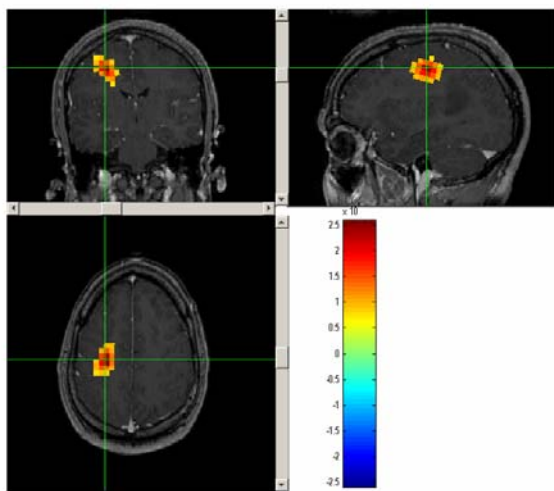


# Hand-motor measurement

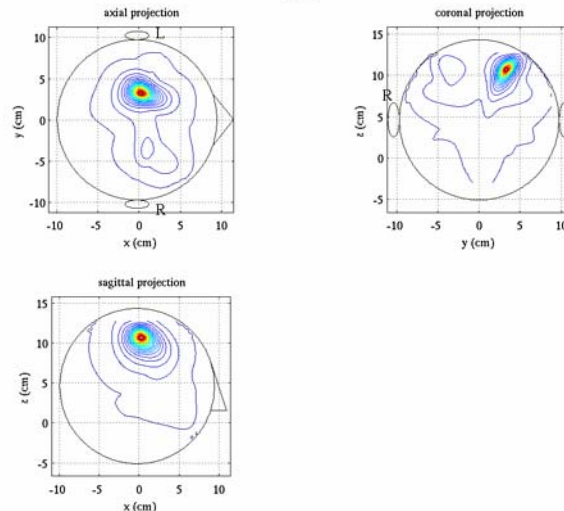


Voluntary right finger movement (every 5 sec)

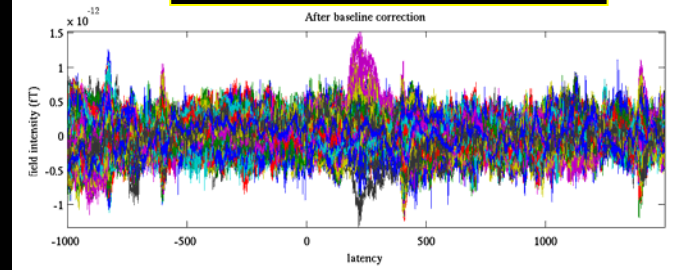
MRI overlay



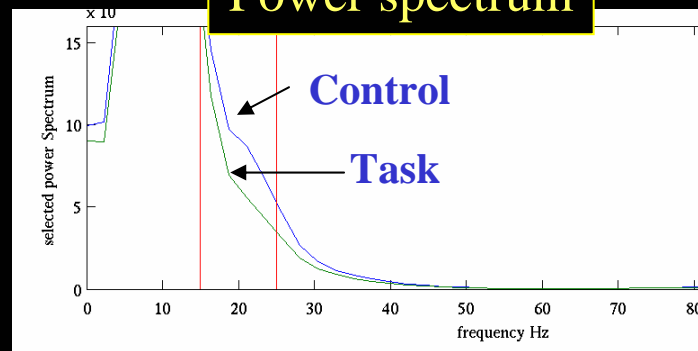
Prewhitening results



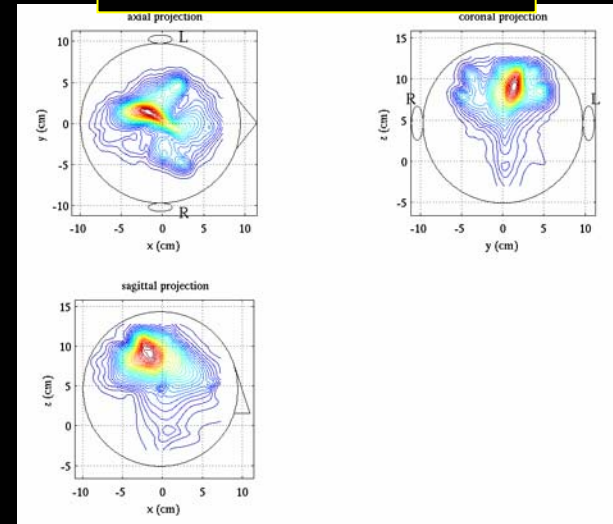
Typical raw epoch



Power spectrum



Subtraction results



# Summary

- We propose a novel prewhitening signal covariance estimation method.
- The proposed prewhitening method is shown to be effective in more realistic (non-ideal) scenarios of two-condition experiments.
- Prewhitening method gives significantly better source reconstruction (less bias and higher spatial resolution) than the existing subtraction-based method.

This investigation is presented at poster P-163,  
(session P4-1) 8/22 3:00—5:00PM

## Collaborators

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Thank you for your attention

Visit <http://www.tmit.ac.jp/~sekihara/>

