

# EEG and MEG source modeling: Beamformer Method

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# Linear Forward/Inverse Problem

Measured data is connected with model parameters via Gain Matrix

$$\mathbf{d} = \underline{\underline{\mathbf{G}}}\mathbf{m}$$

$\mathbf{d}$ : Data vector of size  $N \times 1$

$\mathbf{m}$ : Model parameter vector of size  $M \times 1$

$\underline{\underline{\mathbf{G}}}$ : Gain matrix containing forward calculation with size  $N \times M$

- In case of  $N = M$  there is one unique solution

$$\mathbf{m} = \underline{\underline{\mathbf{G}}}^{-1}\mathbf{d}$$

- For ill-posed problems inversion methods are used to solve equation system
- In non-linear case  $\mathbf{d}$  and  $\mathbf{m}$  are connected via operator  $G(\mathbf{m})$

# What is measured in MEG and EEG?

- Microscopic current flow is generated by neural activity
- Electric potential (EEG) and magnetic field (MEG) at the surface is measured
- Fields can be expressed as a superposition of dipoles
- We want information about:
  - Position
  - Strength
  - Direction



# Potential and Magnetic Field generated by current flow

- EEG and MEG are quasi-static

$$\mathbf{B}(\mathbf{r}) = \frac{\mu_0}{4\pi} \int \mathbf{j}(\mathbf{r}') \times \frac{\mathbf{r} - \mathbf{r}'}{|\mathbf{r} - \mathbf{r}'|^3} d\mathbf{r}'$$

- current density consists of primary current and volume current flow

$$\mathbf{j}(\mathbf{r}') = \mathbf{j}^P(\mathbf{r}') + \mathbf{j}^V(\mathbf{r}') = \mathbf{j}^P(\mathbf{r}') - \sigma(\mathbf{r}') \nabla V(\mathbf{r}')$$

- Forward Problem: Calculate the Potential/Magnetic field at  $\mathbf{r}$  generated by current at point  $\mathbf{r}'$

$$V_0(\mathbf{r}) = \frac{1}{4\pi\sigma_0} \int \mathbf{j}^P(\mathbf{r}') \cdot \frac{\mathbf{r} - \mathbf{r}'}{|\mathbf{r} - \mathbf{r}'|^3} d\mathbf{r}'$$

$$\mathbf{B}_0(\mathbf{r}) = \frac{\mu_0}{4\pi} \int \mathbf{j}^P(\mathbf{r}') \times \frac{\mathbf{r} - \mathbf{r}'}{|\mathbf{r} - \mathbf{r}'|^3} d\mathbf{r}'$$

# Current Dipole

$$\mathbf{j}^P(\mathbf{r}') = \mathbf{q} \delta(\mathbf{r}' - \mathbf{r})$$

with moment  $\mathbf{q} = \int \mathbf{j}^P(\mathbf{r}') d\mathbf{r}'$

- Dipole moment represents neural activity of a source region
- For measurements at the surface different resistivities of brain, skull, scalp has to be considered
- Spherical head model handles this a set of concentric homogeneous shells
- Only radial component away from brain center is measured where volume current terms vanish

$$B_r(\mathbf{r}) = \frac{\mathbf{r}}{r} \cdot \mathbf{B}(\mathbf{r}) = \frac{\mathbf{r}}{r} \cdot \mathbf{B}_0(\mathbf{r}) = \frac{\mu_0}{4\pi} \frac{\mathbf{r} \times \mathbf{r}_q}{r|\mathbf{r} - \mathbf{r}_q|^3} \cdot \mathbf{q}$$

# Algebraic Formulation of Forward Model (Baillet et al. 2001)

- Moment  $\mathbf{q}$  located at  $\mathbf{r}_q$  is separated into dipole magnitude  $q$  and orientation  $\Theta = \{\theta, \varphi\}$
- Electric Potential/Magnetic Field generated by dipole:

$$m(\mathbf{r}) = a(\mathbf{r}, \mathbf{r}_q, \Theta) q$$

where  $a(\mathbf{r}, \mathbf{r}_q, \Theta)$  is the solution to the magnetic or electric forward problem with unit amplitude and orientation  $\Theta$

- Superposition of multiple dipoles yield

$$m(\mathbf{r}) = \sum_i a(\mathbf{r}, \mathbf{r}_{q_i}, \Theta_i) q_i$$

# Algebraic Formulation of Forward Model

$$\begin{aligned} \mathbf{m} &= \begin{bmatrix} m(\mathbf{r}_1) \\ \vdots \\ m(\mathbf{r}_N) \end{bmatrix} = \begin{bmatrix} a(\mathbf{r}_1, \mathbf{r}_{q_1}, \Theta_1) & \cdots & a(\mathbf{r}_1, \mathbf{r}_{q_p}, \Theta_p) \\ \vdots & \ddots & \vdots \\ a(\mathbf{r}_N, \mathbf{r}_{q_1}, \Theta_1) & \cdots & a(\mathbf{r}_N, \mathbf{r}_{q_p}, \Theta_p) \end{bmatrix} \begin{bmatrix} q_1 \\ \vdots \\ q_N \end{bmatrix} \\ &= \underline{\underline{\mathbf{A}}}(\{\mathbf{r}_{q_i}, \Theta_i\}) \mathbf{S}^T \end{aligned}$$

where  $m$  are the EEG/MEG measurements at all  $N$  electrodes,  $\underline{\underline{\mathbf{A}}}$  is the gain matrix relating  $p$  dipoles to the  $N$  sensor locations and  $\mathbf{S}$  contains the source amplitudes

- Each column of  $\underline{\underline{\mathbf{A}}}$  relates one dipole to the sensor array

# Time-dependent Forward Model

$$\underline{\underline{\mathbf{M}}} = \begin{bmatrix} m(\mathbf{r}_1, 1) & \cdots & m(\mathbf{r}_1, T) \\ \vdots & \ddots & \vdots \\ m(\mathbf{r}_N, 1) & \cdots & m(\mathbf{r}_N, T) \end{bmatrix} = \underline{\underline{\mathbf{A}}}(\{\mathbf{r}_{qi}, \Theta_i\}) \begin{bmatrix} \mathbf{s}_1^T \\ \vdots \\ \mathbf{s}_\rho^T \end{bmatrix}$$

for timesteps  $t=1, \dots, T$

# Beamforming as Inversion Method

- Uses spatial filtering to discriminate between signals from region of interest and originating elsewhere
- As the dipole orientation is unknown, we need a set of spatial filters for each cartesian axis denoted as  $\{\Theta_1, \Theta_2, \Theta_3\}$
- Output is a  $3 \times 1$  vector  $\underline{\mathbf{y}} = \underline{\underline{\mathbf{W}}}^T \mathbf{m}$  with a  $3 \times N$  spatial filtering matrix  $\underline{\underline{\mathbf{W}}}^T$
- Spatial filter should filter out all signals that are not generated by the dipole at location  $\mathbf{r}_q$  i. e.

$$\underline{\underline{\mathbf{W}}}^T \underline{\underline{\mathbf{A}}}(\mathbf{r}) = \begin{cases} \underline{\underline{\mathbf{1}}} & |\mathbf{r} - \mathbf{r}_q| \leq \delta : \textit{passband} \\ 0 & |\mathbf{r} - \mathbf{r}_q| > \delta : \textit{stopband} \end{cases}$$

with  $\underline{\underline{\mathbf{A}}}(\mathbf{r}) = [\mathbf{a}(\mathbf{r}, \Theta_1), \mathbf{a}(\mathbf{r}, \Theta_2), \mathbf{a}(\mathbf{r}, \Theta_3)]$  as the  $N \times 3$  forward matrix for 3 orthogonal dipoles at location  $\mathbf{r}$

# Spatial Filtering Matrix

- Problem to solve: Find the weighting coefficients for the filter matrix
- Constraints needed to determine  $\mathbf{W}$
- For EEG/MEG Linearly constrained minimum variance (LCMV) method is commonly used

# LCMV Beamforming

- Nulling the interfering signal weights by minimizing the power of the beamformer subject to a unity gain constraint at  $\mathbf{r}_q$

$$\min_{\underline{\mathbf{W}}^T} \text{tr} \{ \underline{\mathbf{C}}(\mathbf{y}) \} \text{ subject to } \underline{\mathbf{W}}^T \underline{\mathbf{A}}(\mathbf{r}_q) = \mathbf{1}$$

where the covariance matrix  $\underline{\mathbf{C}}(\mathbf{y}) = E[\mathbf{y}\mathbf{y}^T] = \underline{\mathbf{W}}^T \underline{\mathbf{C}}(m) \underline{\mathbf{W}}$  and  $\underline{\mathbf{C}}(m) = E[\mathbf{m}\mathbf{m}^T]$

- Solving this with Lagrange multipliers yields (van Veen et al. 1988):

$$\underline{\mathbf{W}} = \left[ \underline{\mathbf{A}}(\mathbf{r}_q)^T \underline{\mathbf{C}}(m)^{-1} \underline{\mathbf{A}}(\mathbf{r}_q) \right]^{-1} \underline{\mathbf{A}}(\mathbf{r}_q)^T \underline{\mathbf{C}}(m)^{-1}$$

# Estimate Brain Activity

1. Determine Spatial Filter Matrix  $\mathbf{W}$
2. Apply filter to all snapshot vectors  $\mathbf{m}(t)$  for  $t=1, \dots, T$  to estimate dipole moment of a source at  $\mathbf{r}_q$
3. Change location  $\mathbf{r}_q$  and therefore estimate neural activity in different brain regions

# Beamforming: Pro and Contra Factors

## Pro:

- Operate on raw data (no averaging necessary)
- Induced brain activity can be analysed
- Independent from a priori knowledge about number of sources

## Contra:

- Sensitive to constraints
- Sensitive to correlation of activity (cancelling out, partial interfering)
- Sensitive to noise, periods without signal (several methods available to treat noise effects)

Thank you for your attention!

