

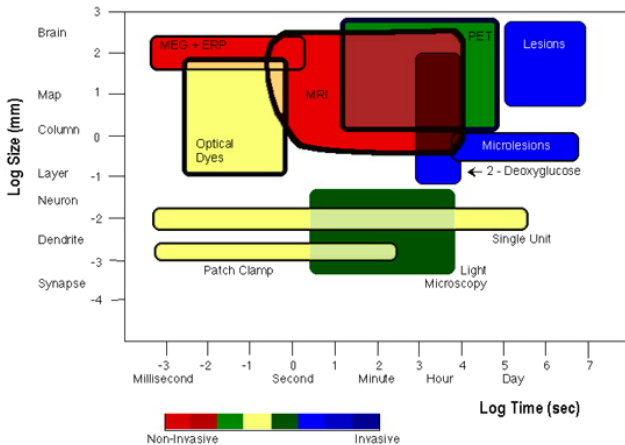
# Estimation of Single Neuron's Current Source Density Distribution by Using Kernel Methods

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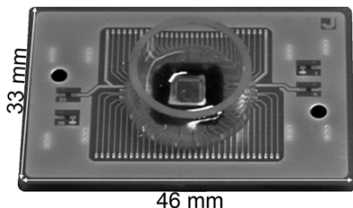
# Spatiotemporal Resolution and Invasiveness of Functional Brain Imaging Methodologies



### APS-MEA III

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Number of Electrodes	4096
Electrode Separation	21 $\mu$ m
Electrode Size	21 $\mu$ m x 21 $\mu$ m
Active Area	2.67mm x 2.67mm
Static Noise	11 $\mu$ Vrms



- ▶ Understanding the neural computational mechanism
- ▶ Information about the inputs
- ▶ Estimation of the ionchannel distribution

Tools: high density  
multielectrode arrays

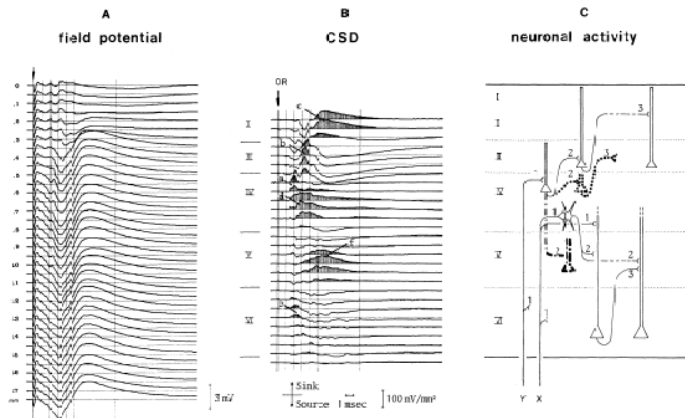
### Goal

Method for estimating the current source density distributions of single cells.

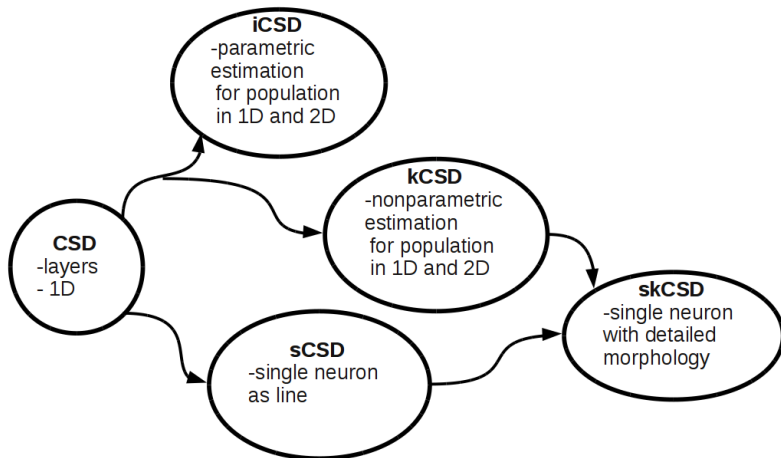
# Current Source Density methods

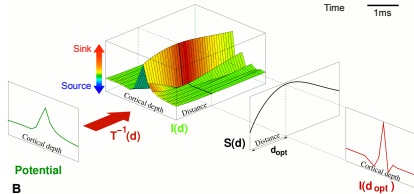
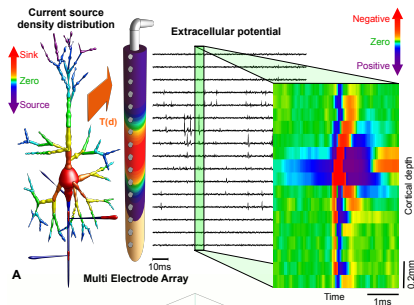
Extracellular potentials  $\Rightarrow$  Currents CSD:

$$C(z_j) = -\sigma \frac{\Phi(z_j + h) - 2\Phi(z_j) + \Phi(z_j - h)}{h^2} \quad (1)$$



# CSD family





Point-source equation:

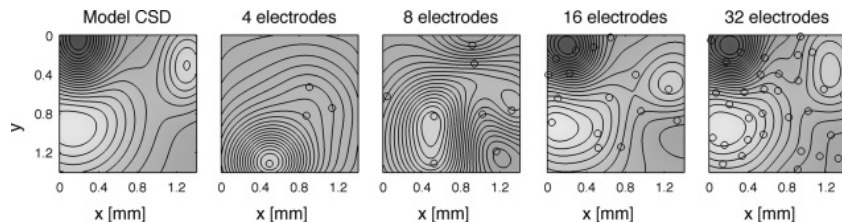
$$\Phi_i(r_i) = \frac{1}{4\pi\sigma} \sum_{j=1}^N \frac{I_j}{|r_i - r_j|} \quad (2)$$

Matrix formalism:

$$\Phi = \mathbf{T}\mathbf{I} \quad (3)$$

Inverse solution:

$$\mathbf{I} = \mathbf{T}^{-1}\Phi \quad (4)$$



- ▶ kernel method
- ▶ number of current sources  $>$  number of electrodes



$\tilde{b}$ : source function - the activity of a neural segment

- Gaussian function

$$\tilde{b}_i(t') = e^{-\frac{(t' - t_i)^2}{R^2}} \quad (5)$$

- Cosinus function

$$\tilde{b}_i(t') = \frac{\cos(|t' - t_i|)\pi}{R}, \quad \text{if } |t' - t_i| < R \quad (6)$$

Current density at  $\mathbf{x}$ :

$$C(\mathbf{x}) = \sum_{j=i}^M a_j \tilde{b}_j(\mathbf{x}) \quad (7)$$

The extracellular potential generated by the  $i$ th source:

$$b_i(x, y, z) = A\tilde{b}_i(t') \quad (8)$$

$$b_i(x, y, z) = \frac{1}{4\pi\sigma} \int \frac{\tilde{b}_i(t')}{\sqrt{(x - x'(t))^2 + (y - y'(t))^2 + (z - z'(t))^2}} dt' \quad (9)$$

$$\Phi(\mathbf{x}) = \sum_{i=1}^M a_i b_i(\mathbf{x}) \quad (10)$$

To determine the CSD distribution in arbitrary positions( $\mathbf{x}$ ), the following kernel functions were introduced:

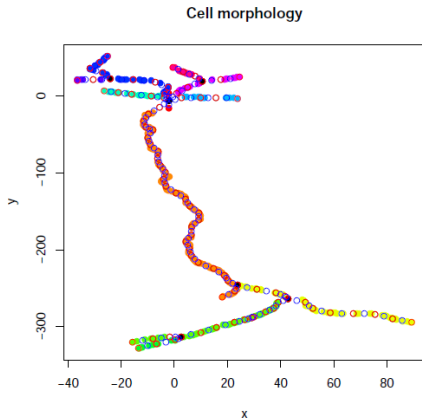
$$K(\mathbf{x}_k, \mathbf{x}_l) = \sum_{i=1}^M b_i(\mathbf{x}_k) b_i(\mathbf{x}_l) = B_k^T B_l \quad (11)$$

$$\tilde{K}(\mathbf{x}_k, \mathbf{y}_l) = \sum_{j=1}^M b(\mathbf{x}_k) \tilde{b}_j(\mathbf{y}_l) = B_k^T \tilde{B}_l \quad (12)$$

Using the simulated or measured extracellular potentials ( $V$ ) and assuming  $\tilde{K}$  is invertible the solution for  $C$  is straightforward.

$$C(\mathbf{x}) = \tilde{\mathbf{K}}^T(\mathbf{x}) \tilde{\mathbf{K}}^{-1} \mathbf{V} \quad (13)$$

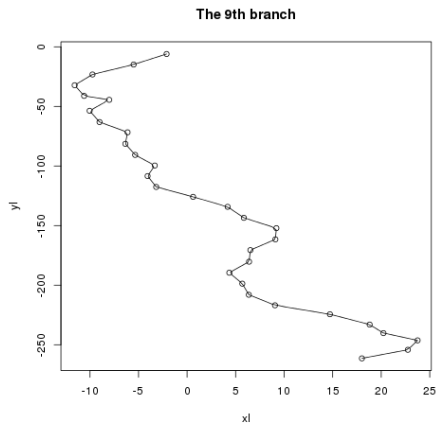
Usually neurons have more, than 1 branch :(



How to handle the branching?

- ▶ branches independent
- ▶ connect the branches

Usually branches are not straight :((



## How to handle?

- ▶ fit a curve
- ▶ position given by parameter  $t$

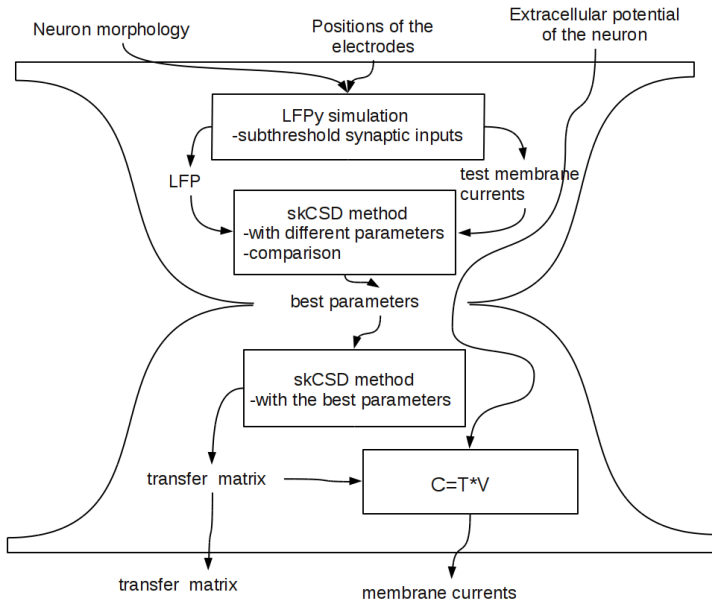
# Lots of parameters :(((

Parameters of the method:

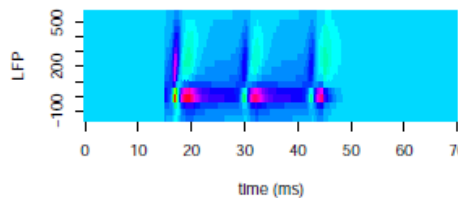
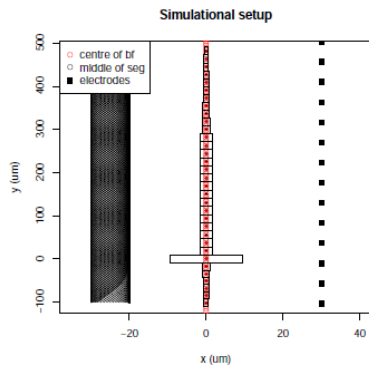
- ▶ type of basis function
- ▶ number of basis function
- ▶ width of basis function

Parameters of the test simulations:

- ▶ neuron morphology
- ▶ inputs
- ▶ cell to electrode distance
- ▶ position of the electrode (1D, 2D, 3D)



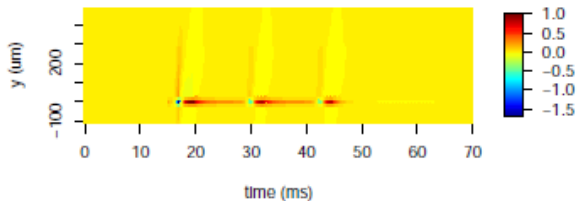
# Ballstick neuron



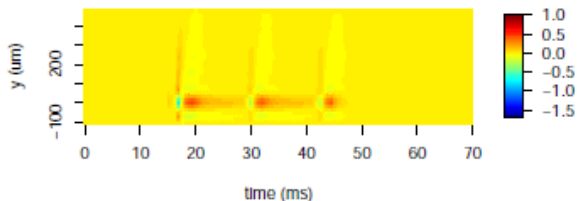


# ksCSD for ballstick neuron

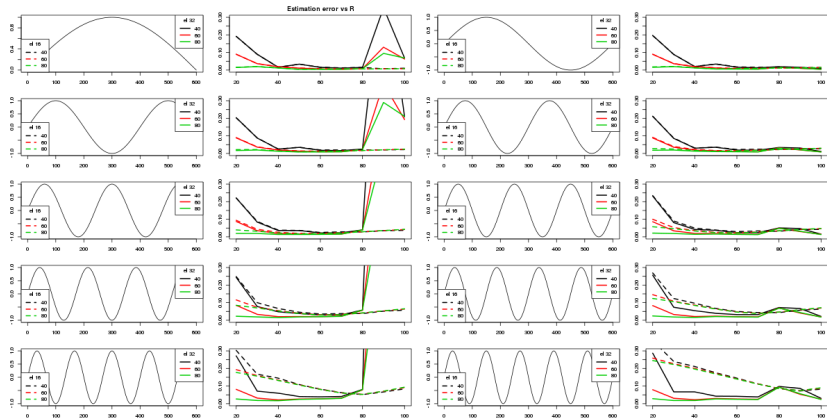
Membrane currents



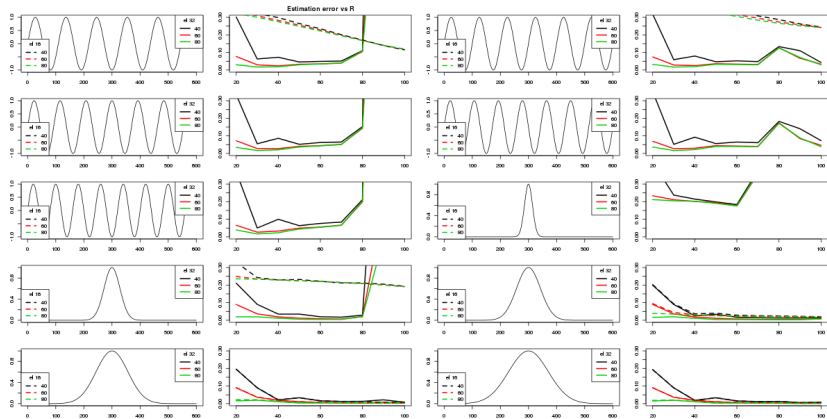
Estimated MC



# Ballstick neuron

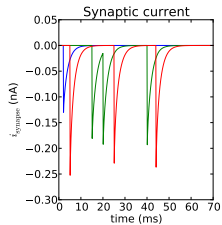
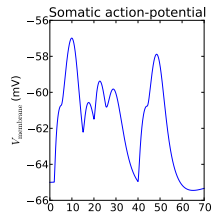
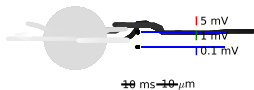


# Ballstick neuron

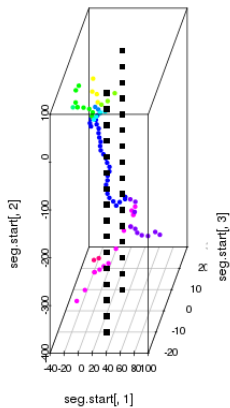


# Complex morphology

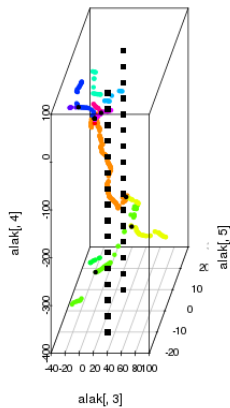
Location-dependent extracellular spike shapes



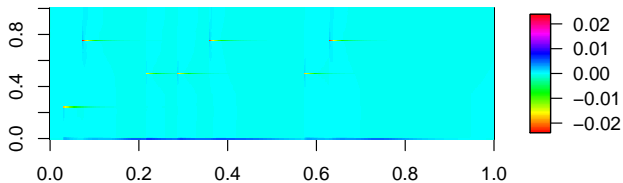
The cell and the electrode



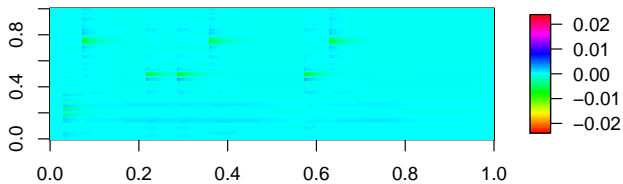
The cell and the electrode

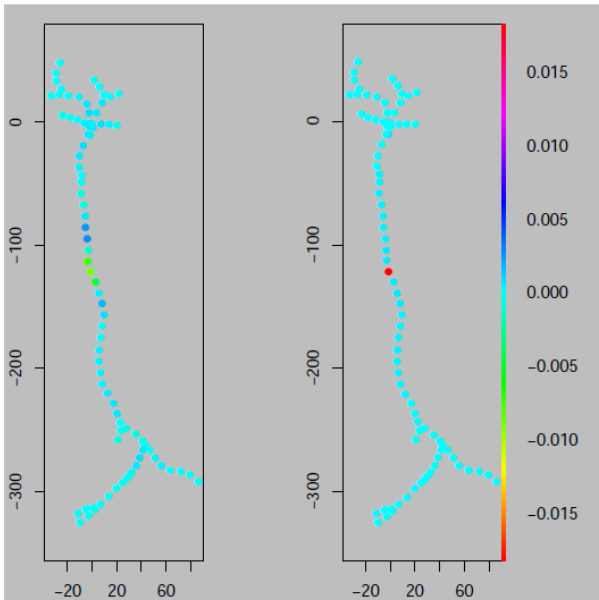


**Original**



**ksCSD**





# Future plans

- ▶ Run more test simulations
- ▶ Make the program usable also for others (GUI)
- ▶ Test the method on experimental data



Thanks for the attention!

Which is the best set of parameters?

$$e = \frac{\sum_{t,i} |C_{skCSD} - C_o|}{\sum_{t,i} |C_o|} \quad (14)$$