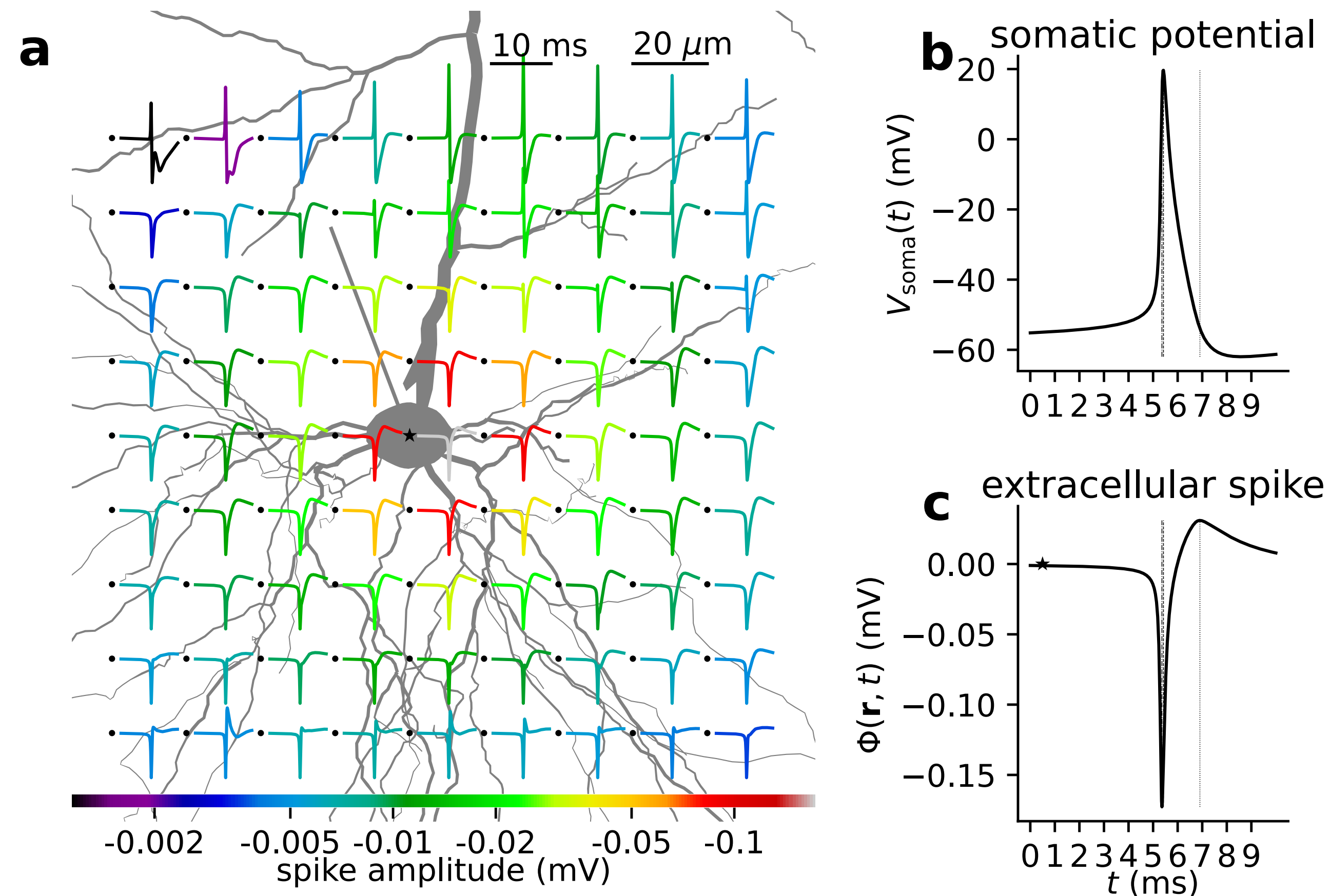


# LFPy & related tools (CodeJam #12)

Espen Hagen, Norwegian University of Life Sciences (NMBU)  
[espen.hagen@nmbu.no](mailto:espen.hagen@nmbu.no)

# LFPy

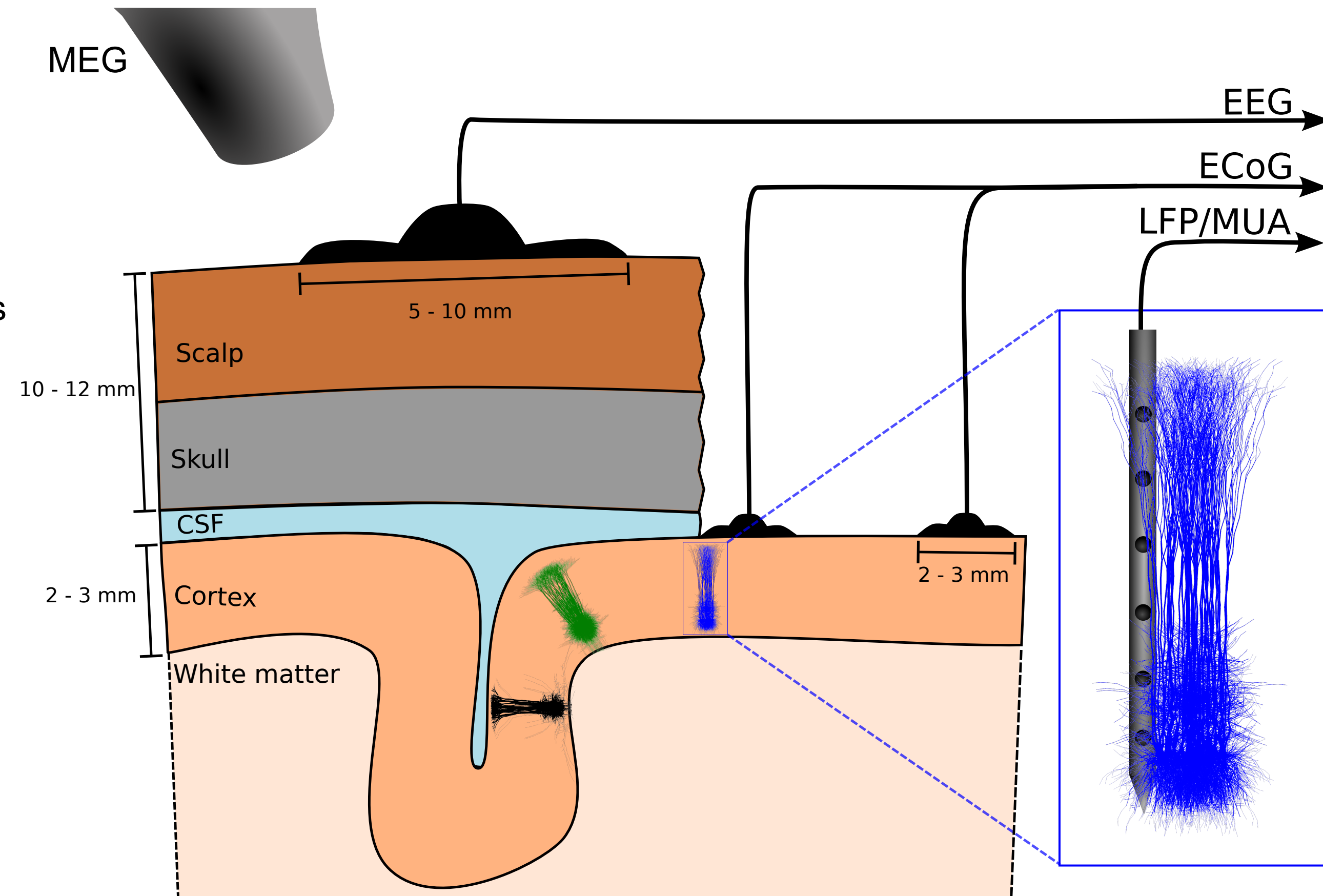
- Python package for calculation of extracellular signals from multicompartment (MC) neuron models and networks
  - Key neural simulators focus on intracellular dynamics ( $V_m$  etc.), network dynamics (spikes)
  - Much insight from extracellular electrophysiological recordings of neural activity (spikes, LFP, ECoG, EEG, MEG)
  - Uses NEURON ([neuron.yale.edu](http://neuron.yale.edu)) under the hood
- Sources: [github.com/LFPy/LFPy](https://github.com/LFPy/LFPy)
- Docs: [lfp.py.rtdf.io](http://lfp.py.rtdf.io)
- Installation:
  - `pip install LFPy`
  - `conda install lfp.py -c conda-forge`
- Collab w. example notebooks: [wiki.ebrains.eu/bin/view/Collabs/lfp.py-showcase](http://wiki.ebrains.eu/bin/view/Collabs/lfp.py-showcase)



Action-potential waveforms from L5 PC model (Lindén et al. (2014), *Front NeuroInform* 7:41)

# LFPy

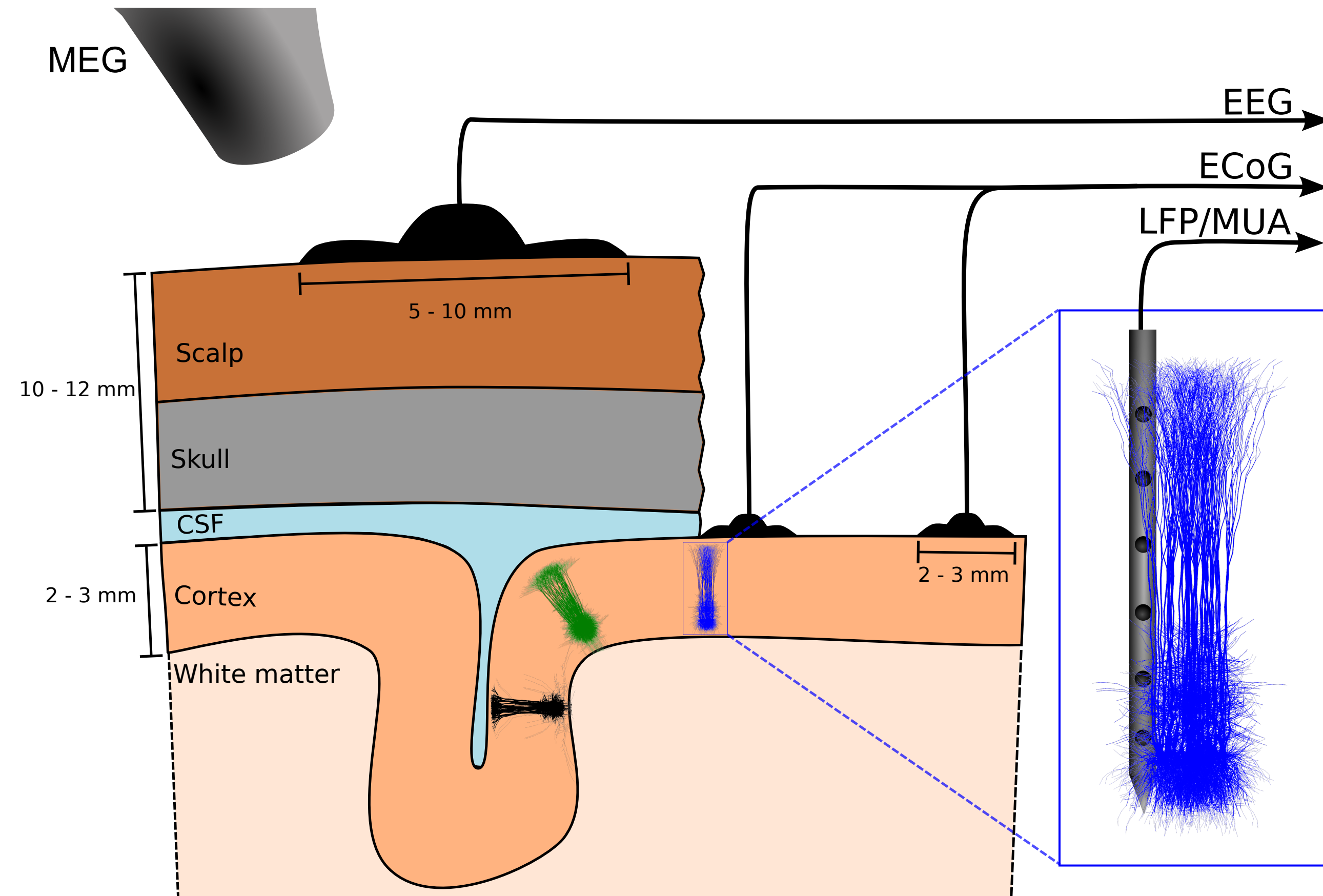
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Hagen et al. (2018), *Front NeuroInform* 12:92

# LFPy

- High-level class objects representing:
  - cells
  - synapses
  - intracellular stimulation devices
  - extracellular recording devices
  - networks
- Publications:
  - Lindén H, Hagen E, Łęski S, Norheim ES, Pettersen KH and Einevoll GT (2014) *Front. Neuroinform.* 7:41. doi: [10.3389/fninf.2013.00041](https://doi.org/10.3389/fninf.2013.00041)
  - Hagen E, Næss S, Ness TV and Einevoll GT (2018) *Front. Neuroinform.* 12:92. doi: [10.3389/fninf.2018.00092](https://doi.org/10.3389/fninf.2018.00092)



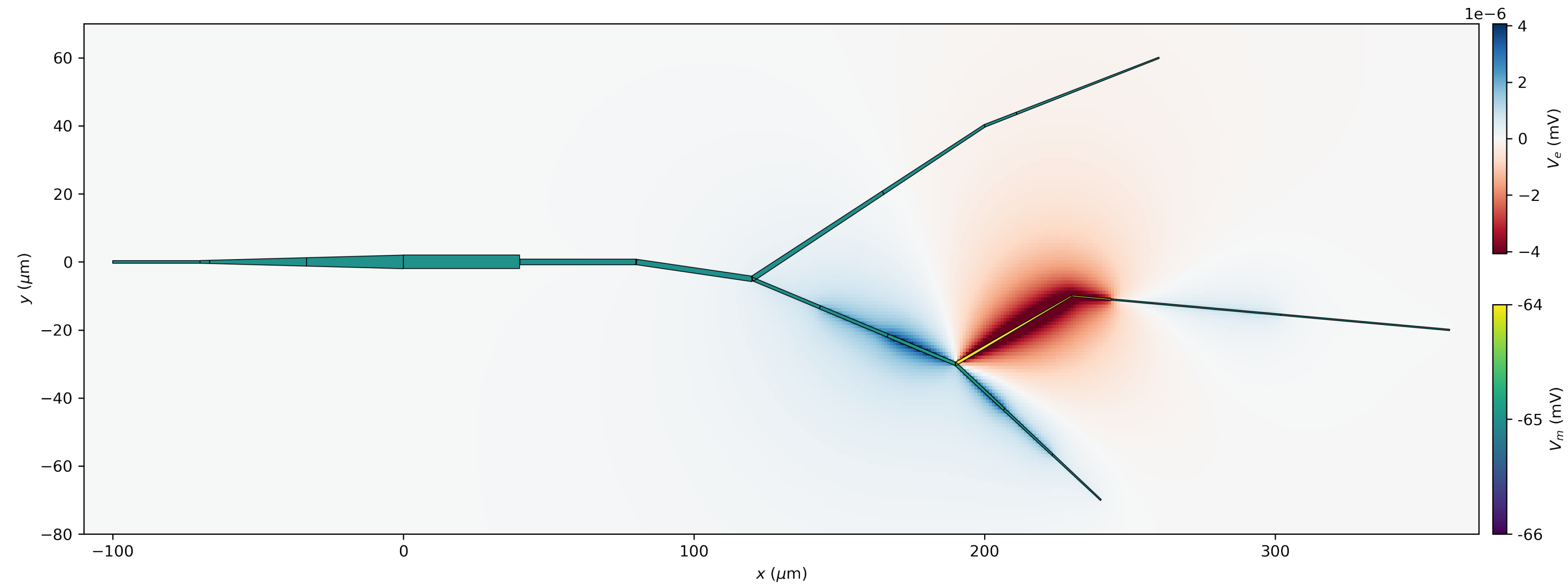
Hagen et al. (2018), *Front NeuroInform* 12:92

# LFPykit

- Simulator-independent forward models derived using linear volume-conductor theory
- Linear mapping between transmembrane currents and different signals

- $\psi(\vec{R}, t) = \mathcal{F}\mathbf{I}_m(\vec{r}, t)$

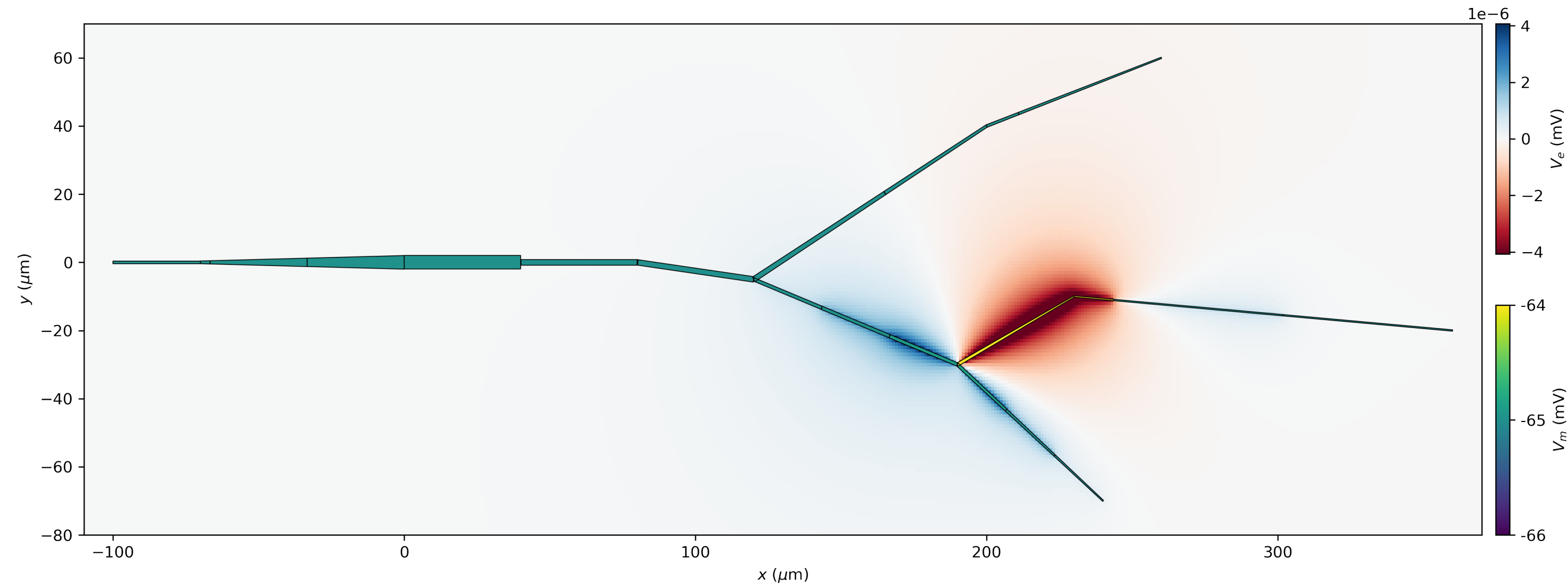
- point/line sources (infinite homogeneous/inhomogeneous/anisotropic media)
- current source density
- current dipole moment
- EEG signals (idealised 4-sphere head model or detailed NY head model)
- MEG signals (infinite homogeneous or spherically symmetric conductor)



Extracellular potentials from few-compartment model implemented in Arbor and line-source approximation — `lfpykit.LineSourcePotential` (LFPykit/examples/Example\_Arbor\_swc.ipynb)

# LFPykit

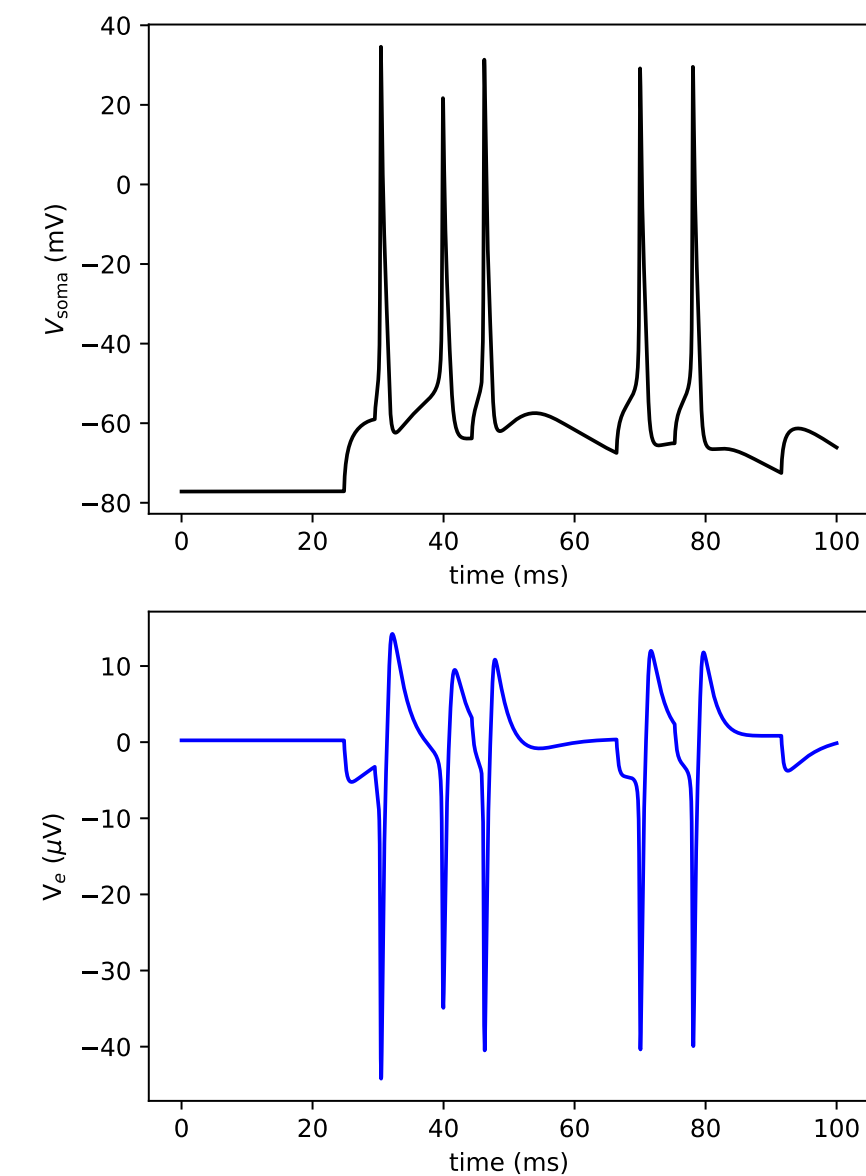
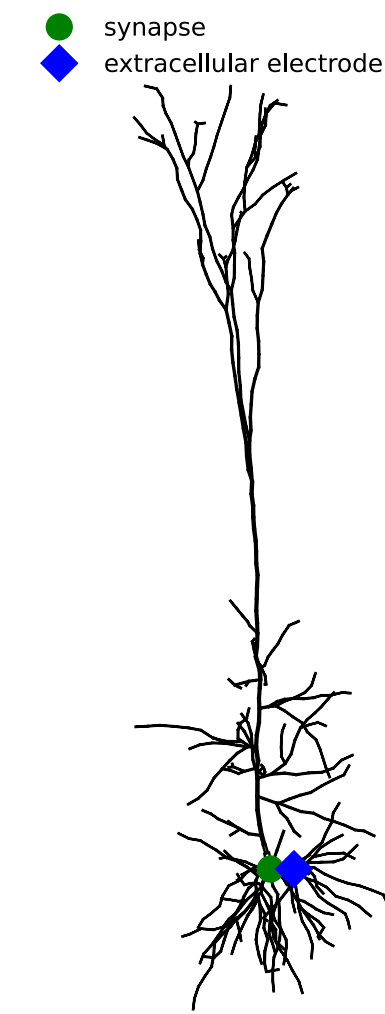
- Dependency in LFPy itself
- Usage examples with plain NEURON, Arbor ([arbor-sim.org](http://arbor-sim.org)) and LFPy
- Sources: [github.com/LFPy/LFPykit](https://github.com/LFPy/LFPykit)
- Docs: [LFPykit.rtf.d.io](http://LFPykit.rtf.d.io)
- Installation:
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  - `conda install lfpypykit -c conda-forge`
- Collab w. example notebooks: [wiki.ebrains.eu/bin/view/Collabs/lfpypykit-showcase/](http://wiki.ebrains.eu/bin/view/Collabs/lfpypykit-showcase/)



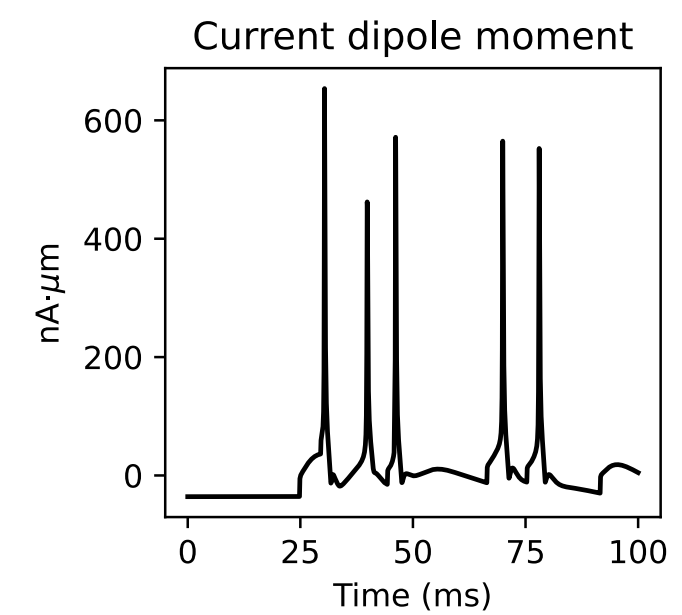
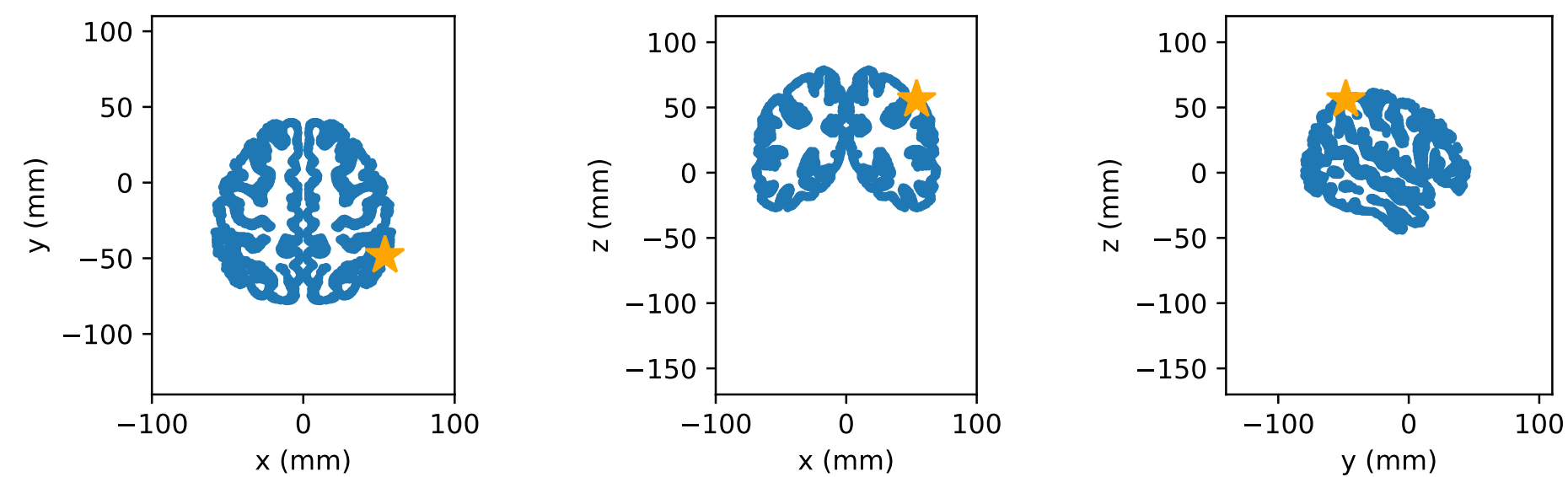
Extracellular potentials from few-compartment model implemented in Arbor and line-source approximation — `lfpypykit.LineSourcePotential` (LFPykit/examples/Example\_Arbor\_swc.ipynb)

# LFPykit

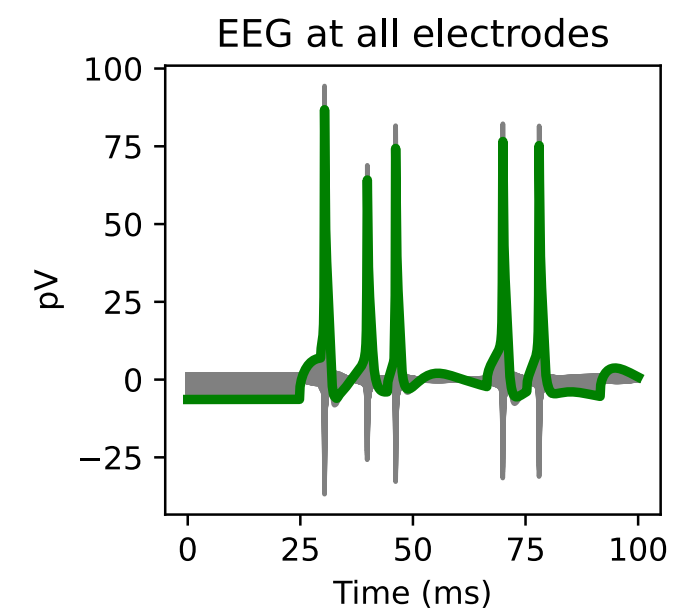
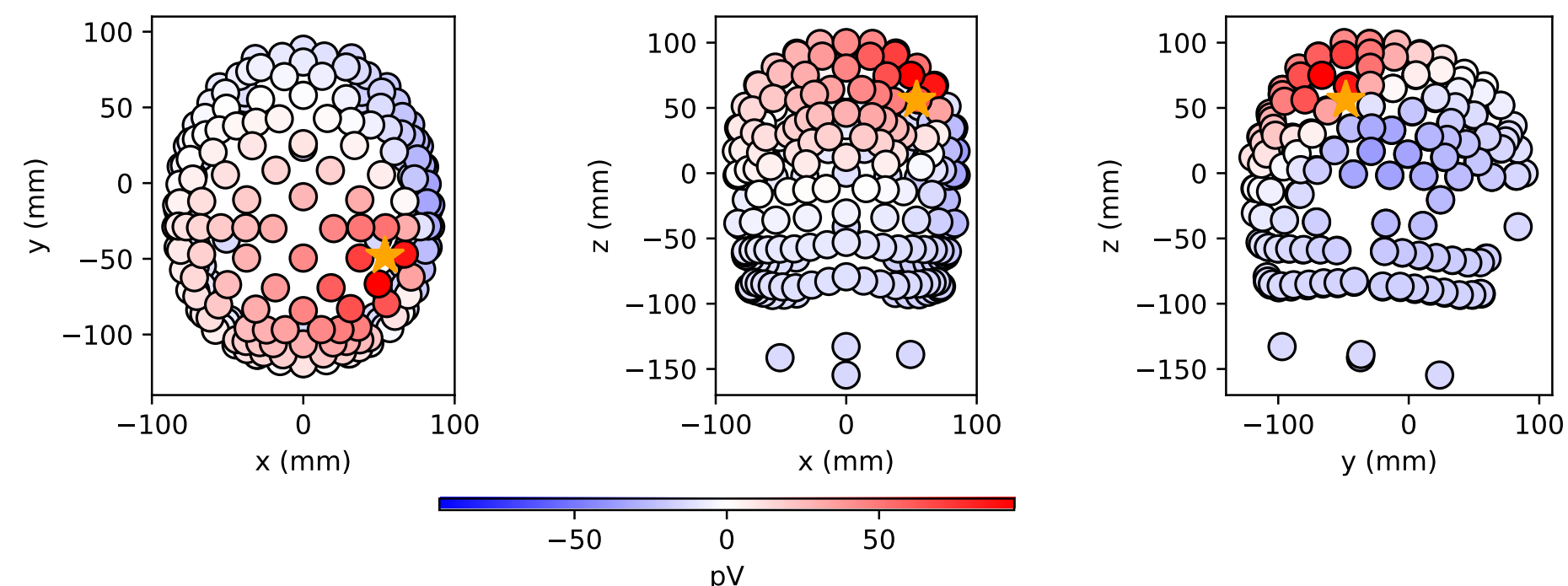
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Cortex



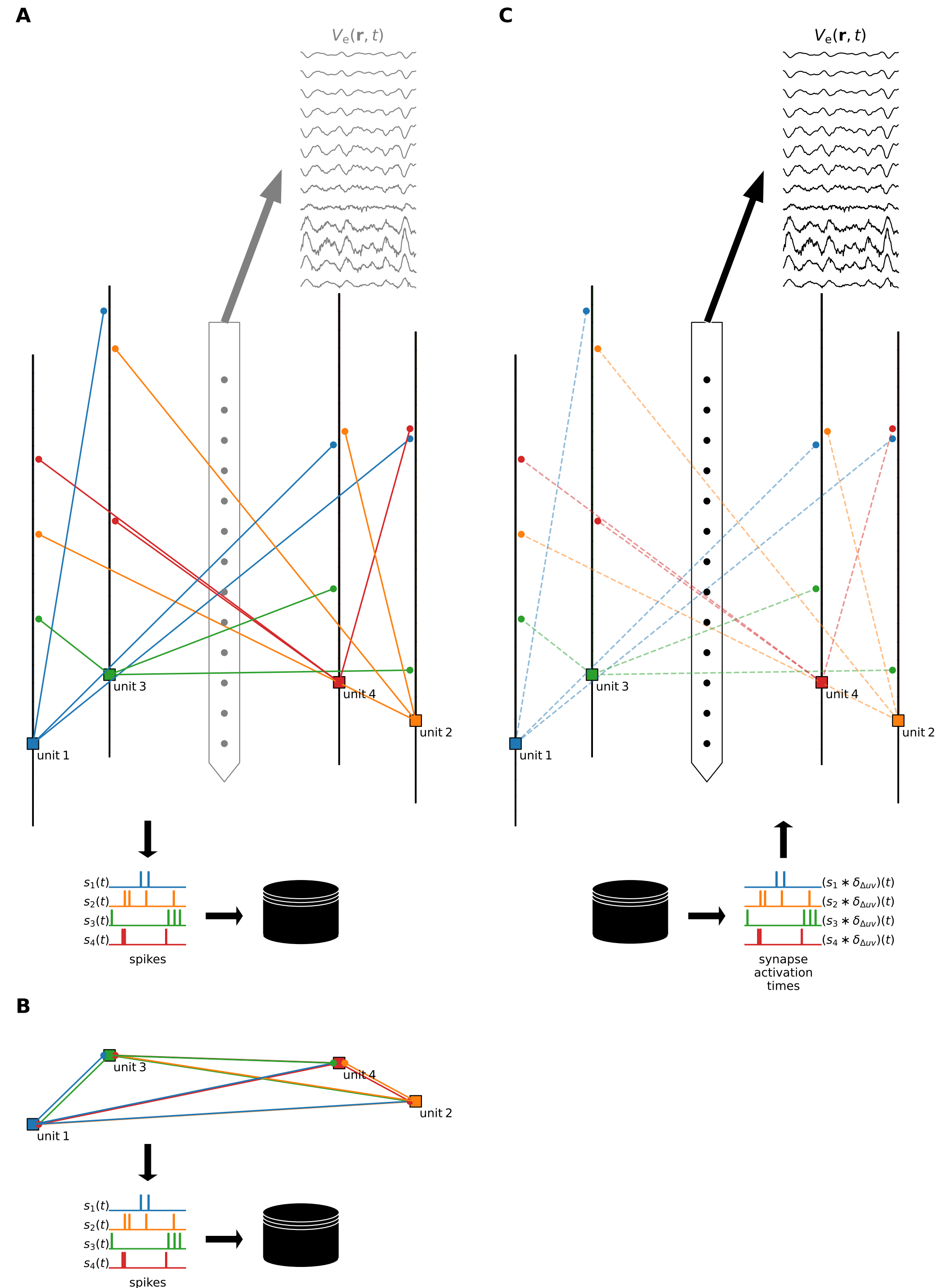
EEG



EEG scalp potentials from 'NY Head' model (Huang et al. (2016), *NeuroImage* 140) from current dipole moment using LFPy, `lfpykit.CurrentDipoleMoment` & `lfpykit.eegmegcalc.NYHeadModel` (LFPykit/examples/LFPykit\_demo.ipynb)

# hybridLFPy

- Hybrid scheme (Hagen et al. (2016), *Cereb Cortex* 26:12):
  - disentangles predictions of network activity (spiking) from FM predictions
  - simulations of spiking using simplified (point-neuron) networks
  - MC neurons for FM predictions
- Sources: [github.com/INM-6/hybridLFPy](https://github.com/INM-6/hybridLFPy)
- Docs: [hybridlfp.py.rtf.d.io](https://hybridlfp.py.rtf.d.io)
- Installation: `pip install hybridLFPy`
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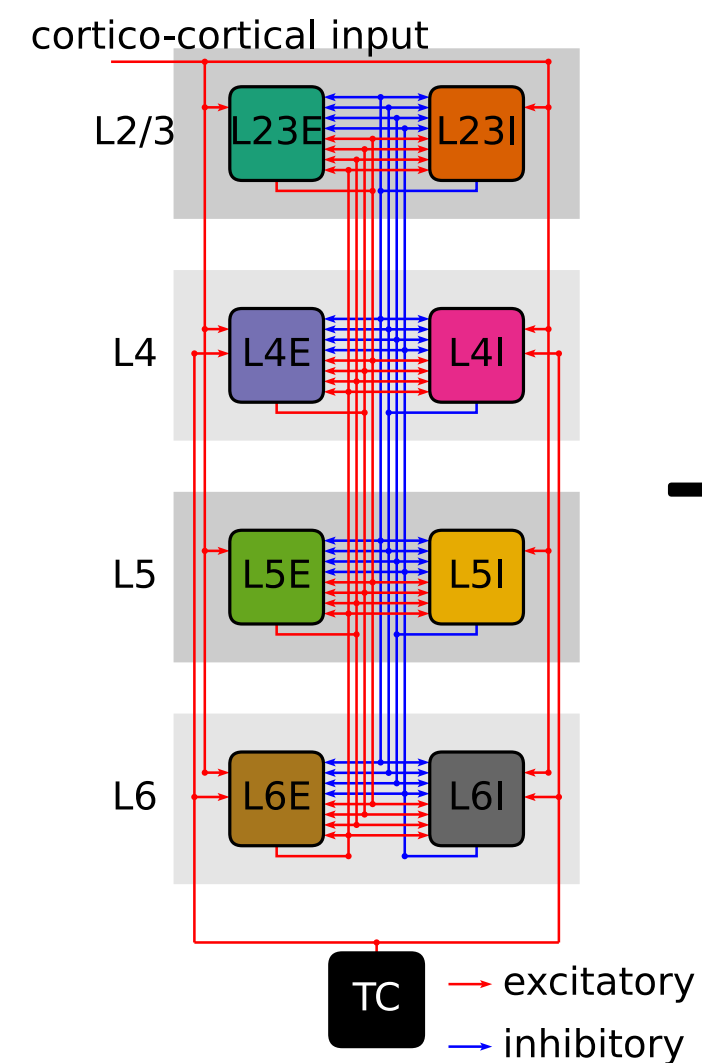


# hybridLFPy

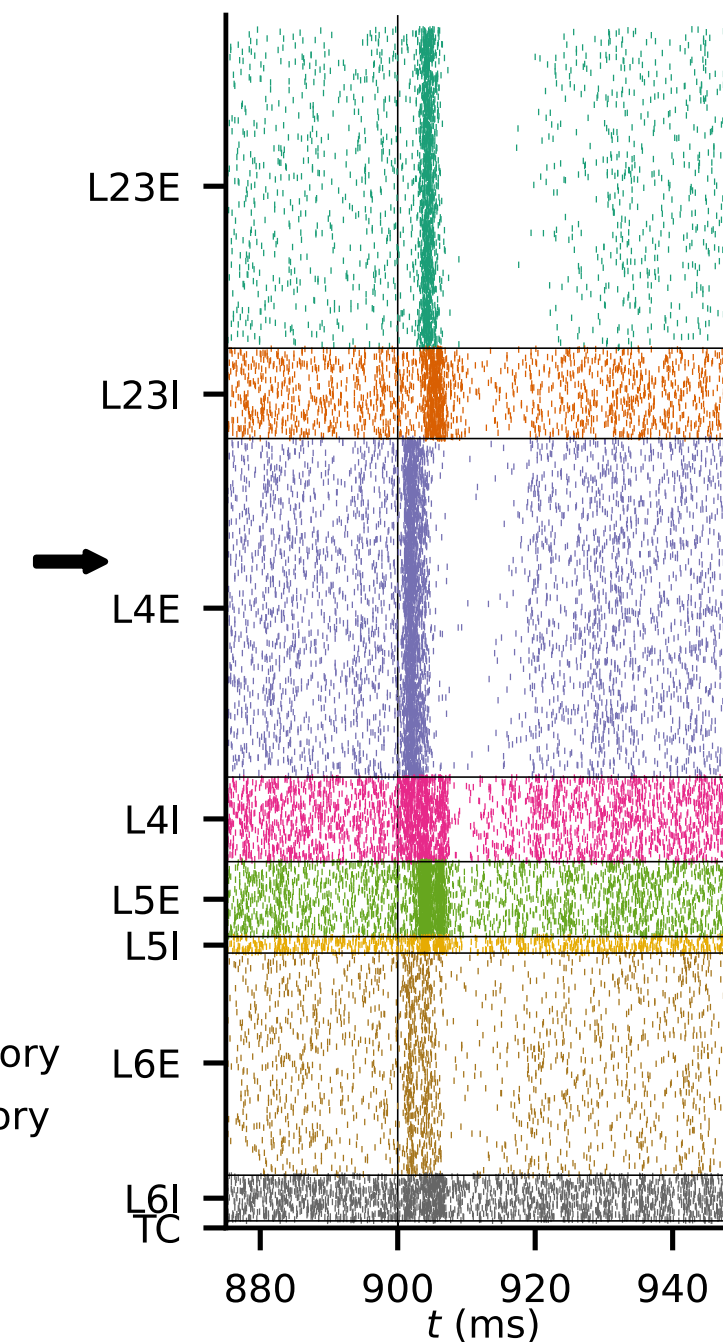
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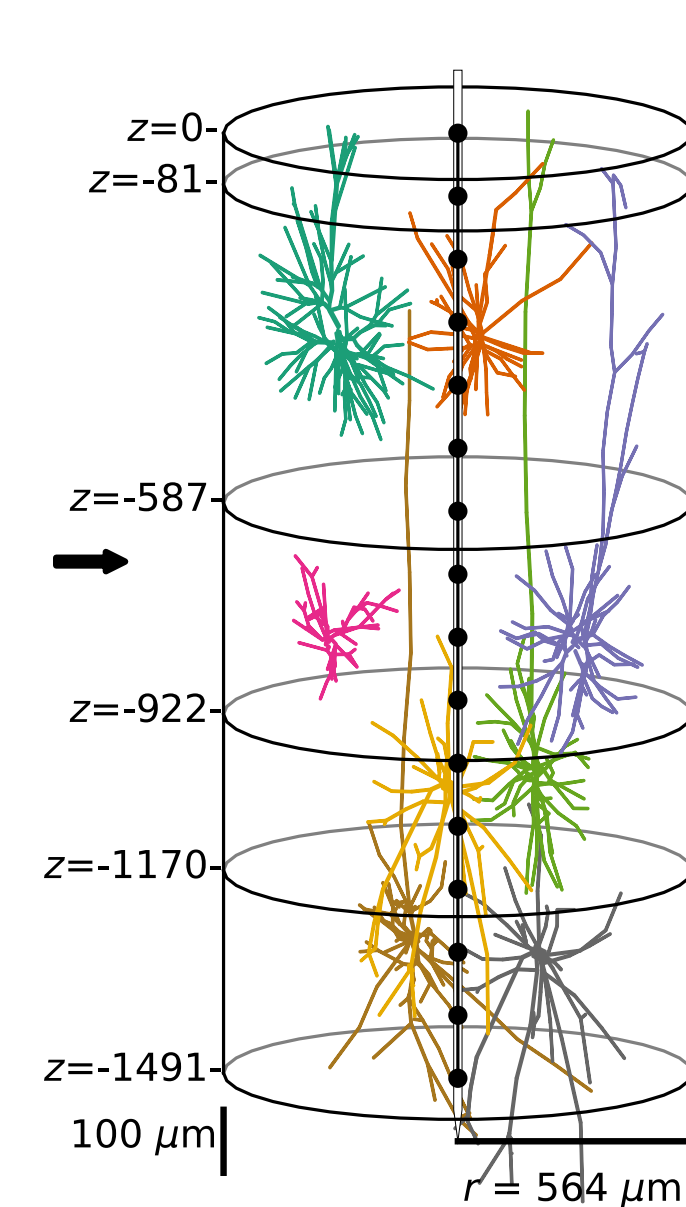
**A** point-neuron network



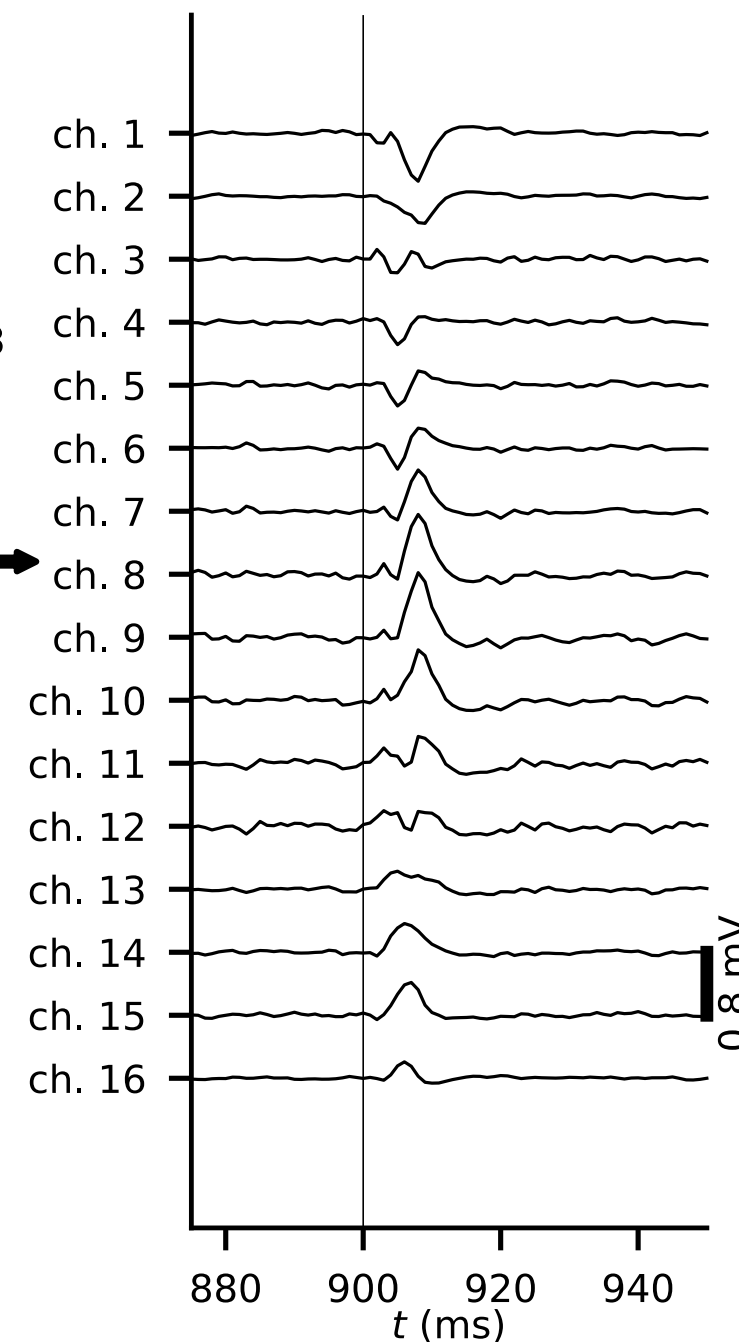
**B** spiking activity



**C** multicompartiment neurons



**D** LFP



LFP predictions from cortical microcircuit model by Potjans & Diesmann (2014, *Cereb Cortex* 24:3) in Hagen et al. (2016, *Cereb Cortex* 26:12)

# LFPy kernels

- FM based calculations of causal spike-signal impulse response functions  $H_{YX}(\vec{r}, \tau)$  for finite-sized neuronal network models

- Extracellular signal approximation:  

$$\psi(\vec{r}, t) = \sum_X \sum_Y (\nu_X * H_{YX})(\vec{r}, t)$$

- One MC neuron simulation required per pathway between pre-synaptic population  $X$  and postsynaptic population  $Y$

- Intended for spike- and rate-based frameworks

- Sources: [github.com/LFPy/LFPy kernels](https://github.com/LFPy/LFPy kernels)

- Docs: [lfpynotes.github.io](https://lfpynotes.github.io/)

- Installation:

- `pip install LFPy kernels --pre`

- Collab w. example notebooks (coming):  
[wiki.ebrains.eu/bin/view/Collabs/lfpynotes](https://wiki.ebrains.eu/bin/view/Collabs/lfpynotes)

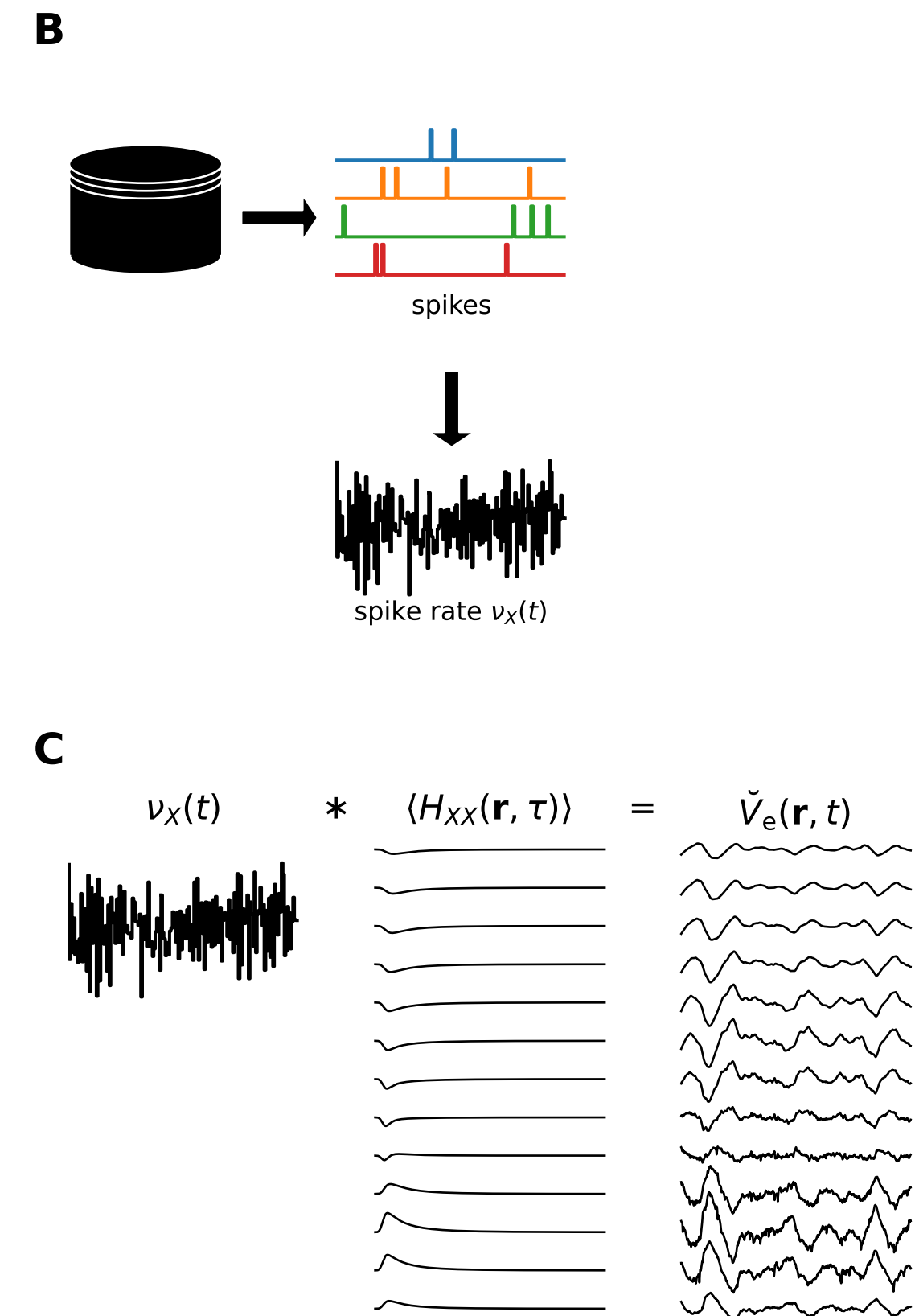
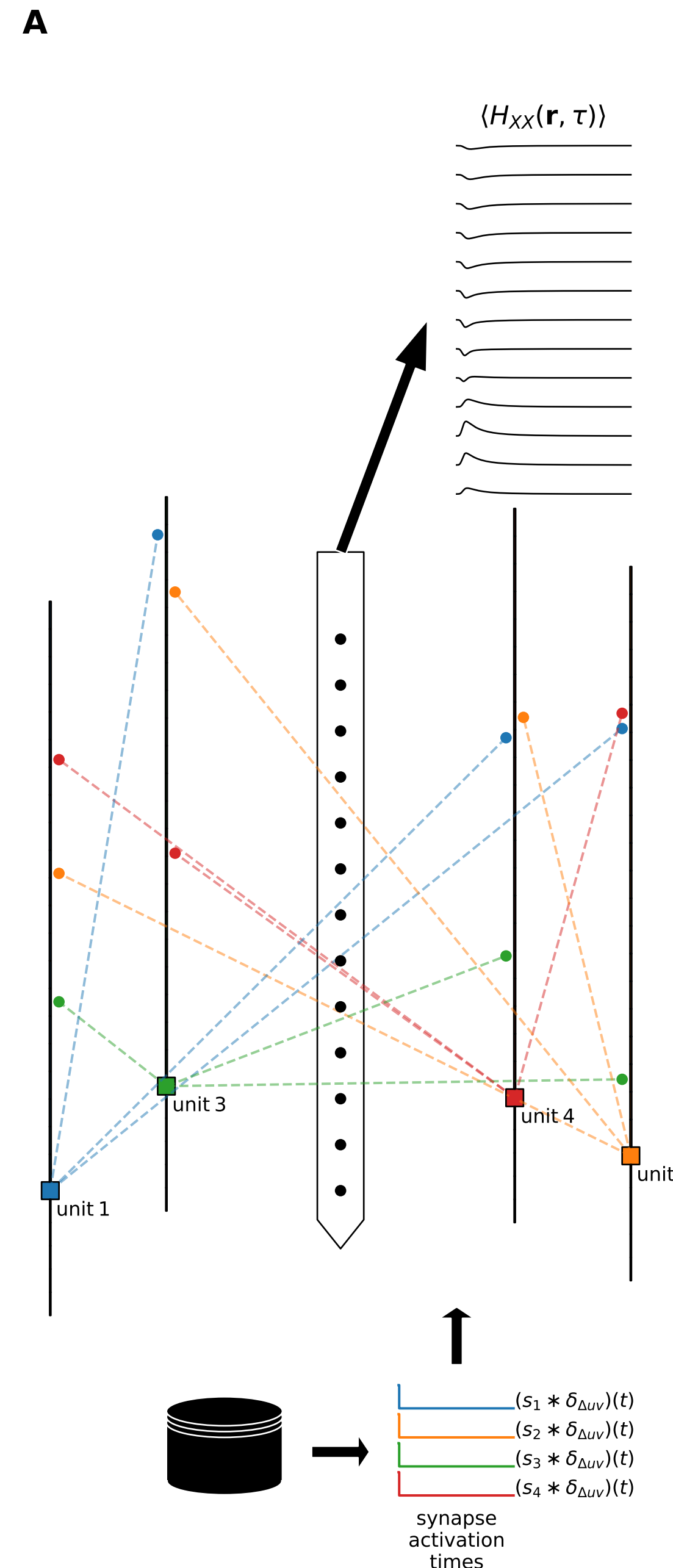


Illustration of 'hybrid scheme' predictions of population-averaged spike-LFP 'kernels' and signal reconstruction convolving spike rates with corresponding kernels

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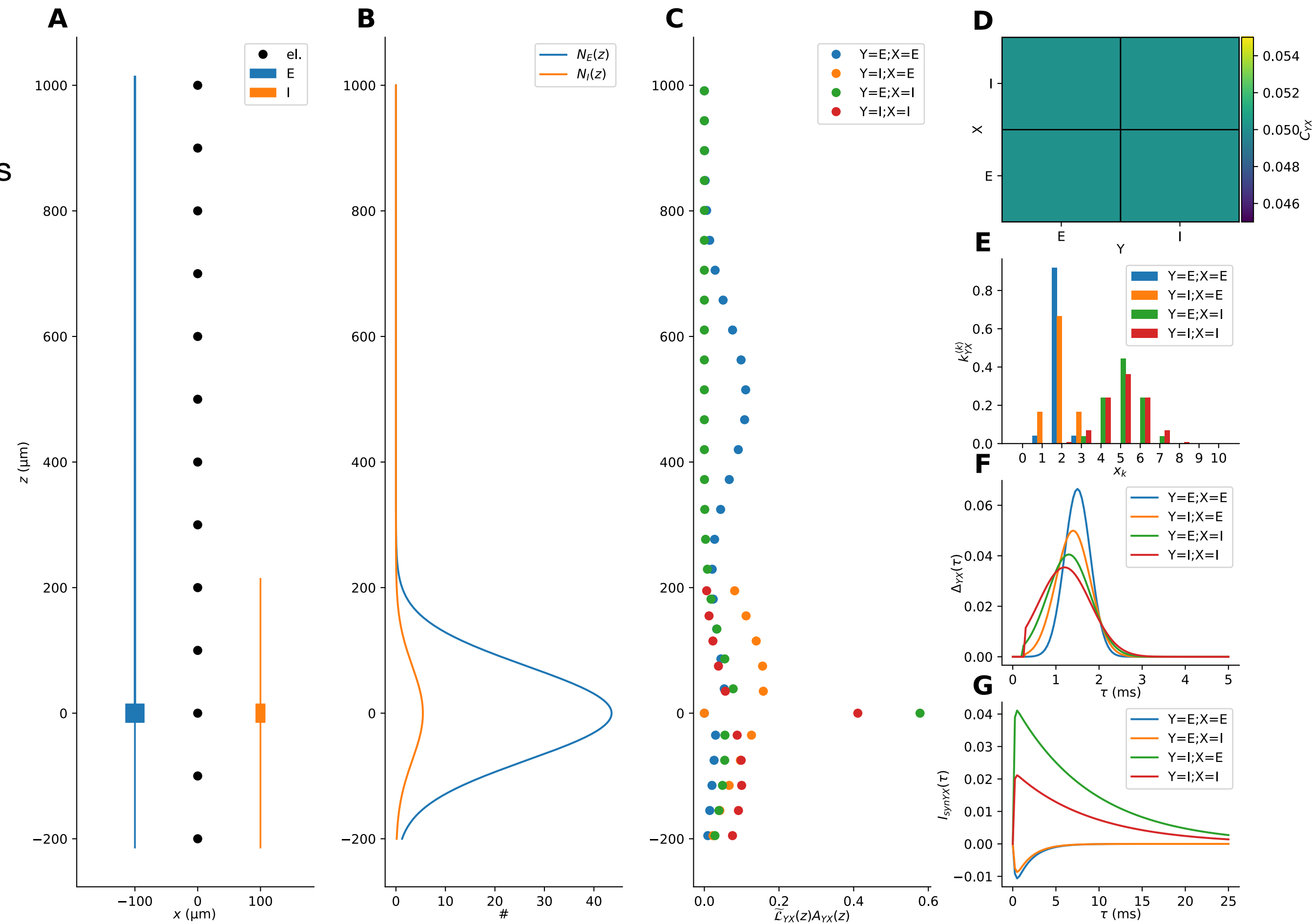
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Components for deterministic kernel predictions for finite-sized 2-population networks, using linearised cell models, distributions of cells, connection probabilities, multipulse counts, delays, linearised synapse currents.

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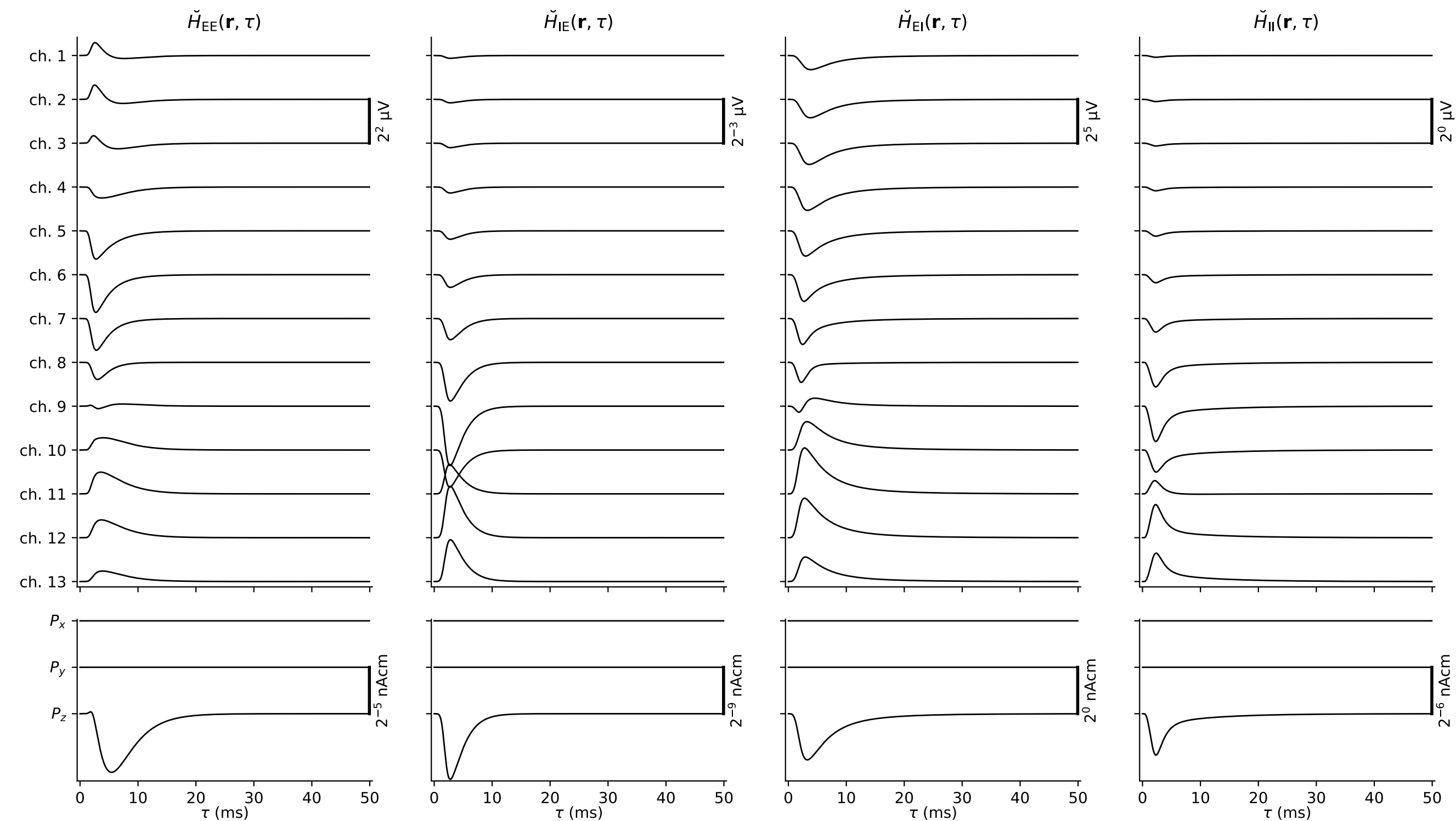
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- Docs: [lfpykernels.rtd.io](https://lfpykernels.rtd.io)

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Kernels for each connection pathway in two-population network presynaptic population spikes to LFPs and current dipole moments  
(LFPykernels/examples/LIF\_net\_forward\_model\_predictions.ipynb)

# LFPy kernels

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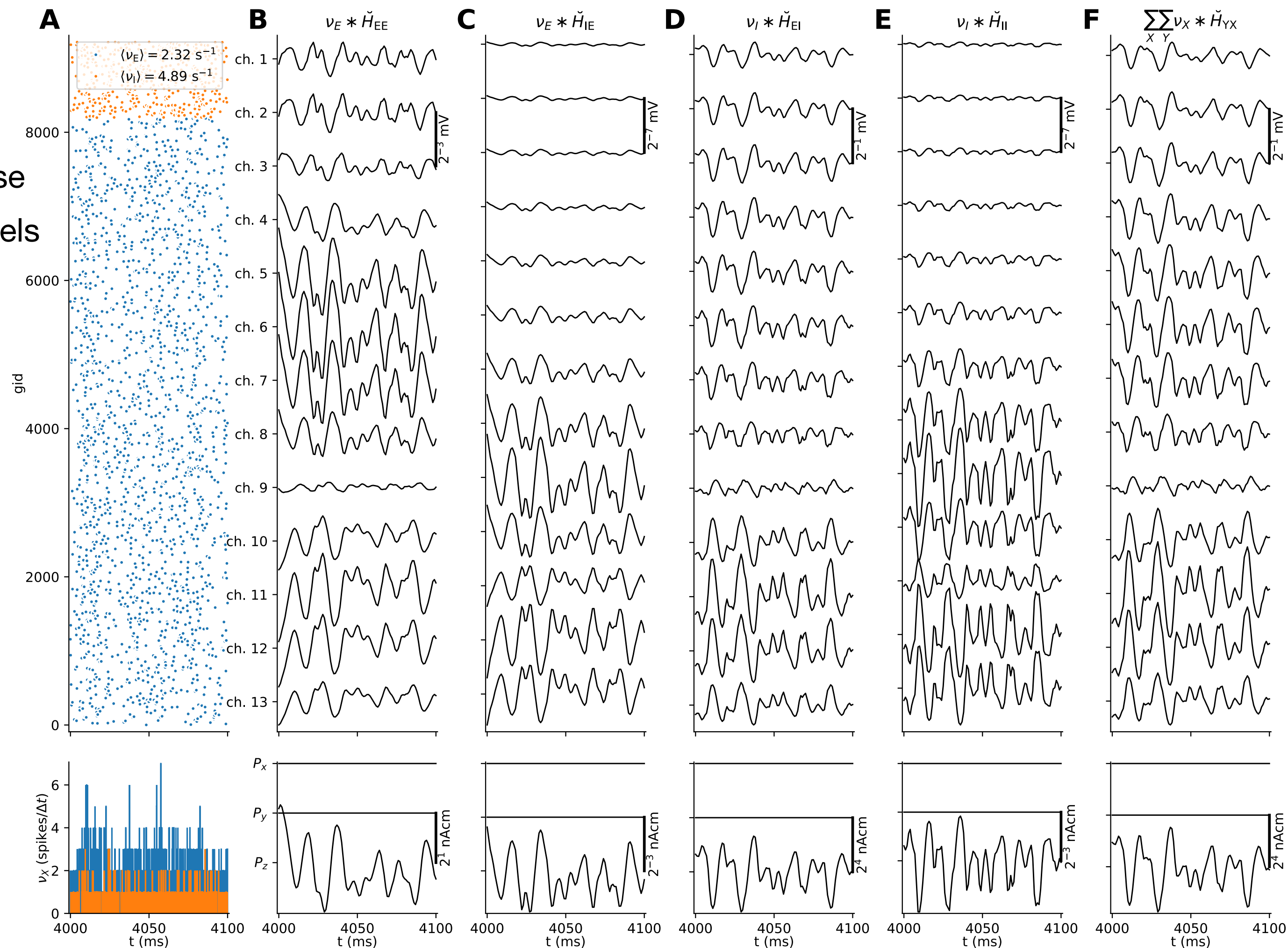
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Application of kernels with two-population point-neuron network in NEST. Signals computed on the fly using FIR\_filter.nestml (LFPy kernels/examples/LIF\_net\_forward\_model\_predictions.ipynb)

**Questions?**