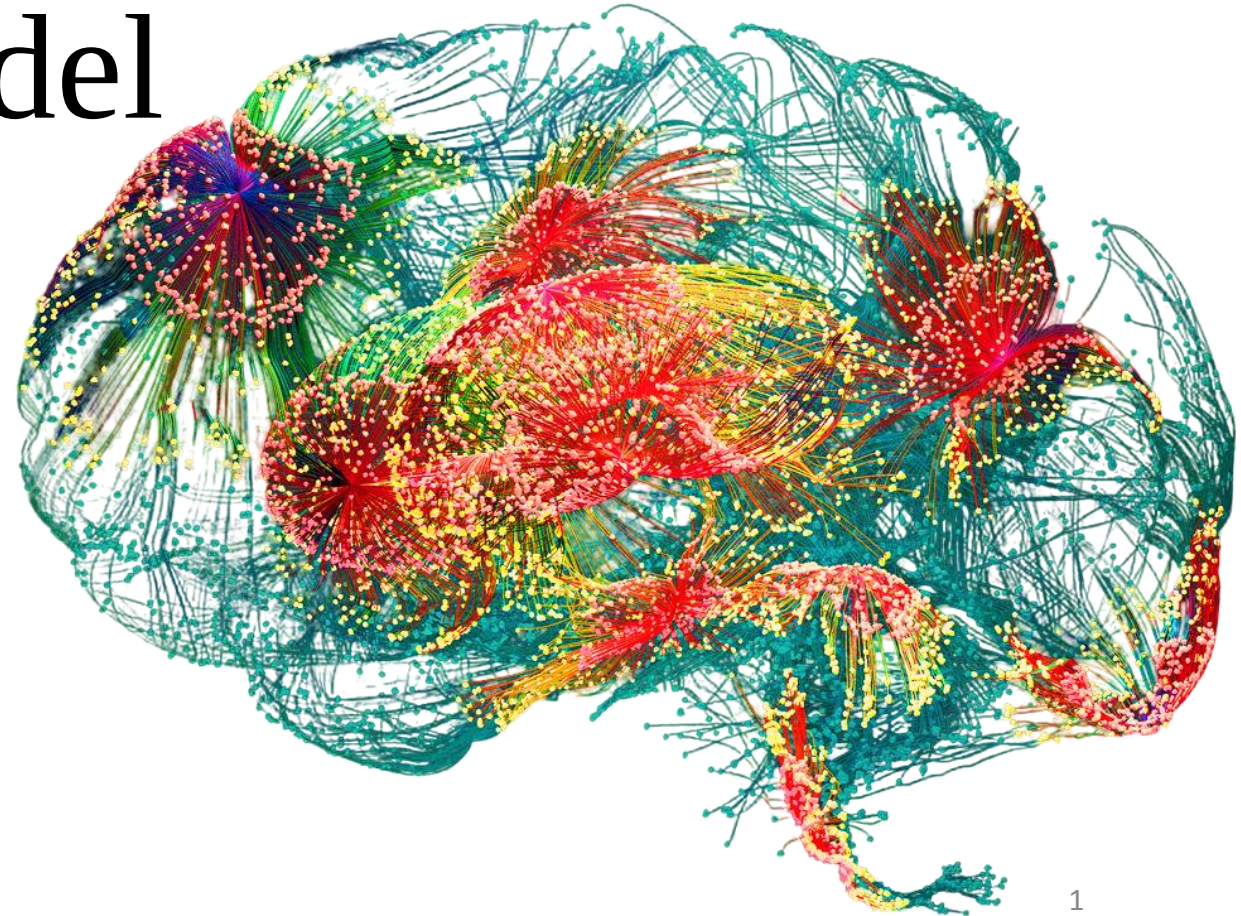


Simulating EEG Signals Using Wilson-Cowan based coupled Brain Network Model



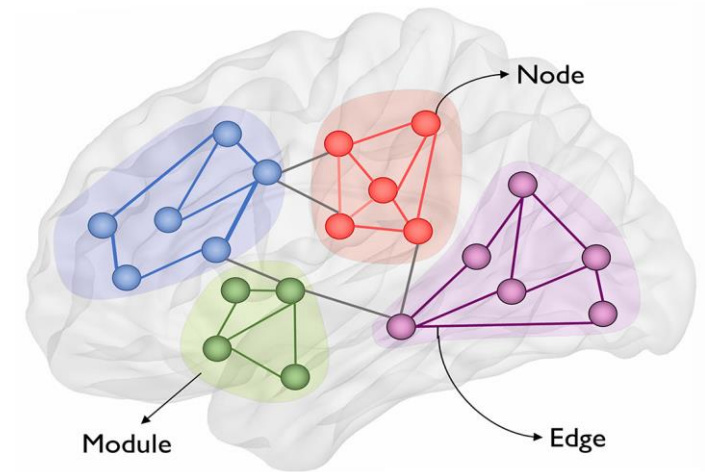
Brains, Dynamics & Computation
Workshop on Network Neuroscience
June 04, 2025

Radhika Shrimankar , Amrita Mind Brain Centre, Kerala
Nimish Kumar, IIT Jodhpur
Sneha Biswas, NITK Surathkal

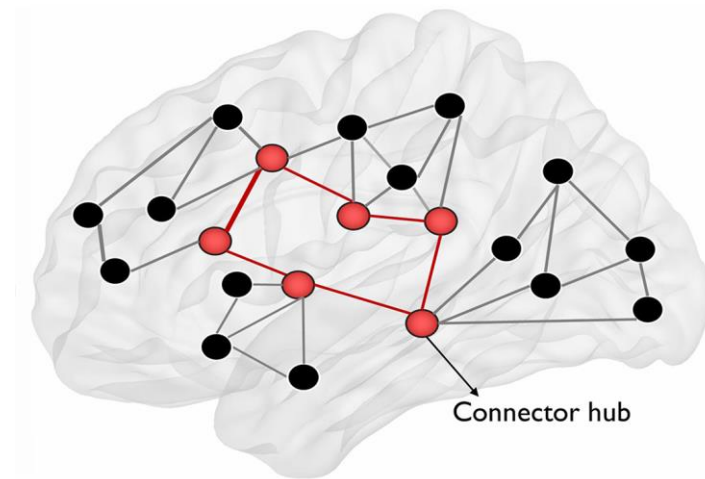


Brain as a Network

- Complex Connectivity
- Nodes
- Structural vs Functional Connectivity



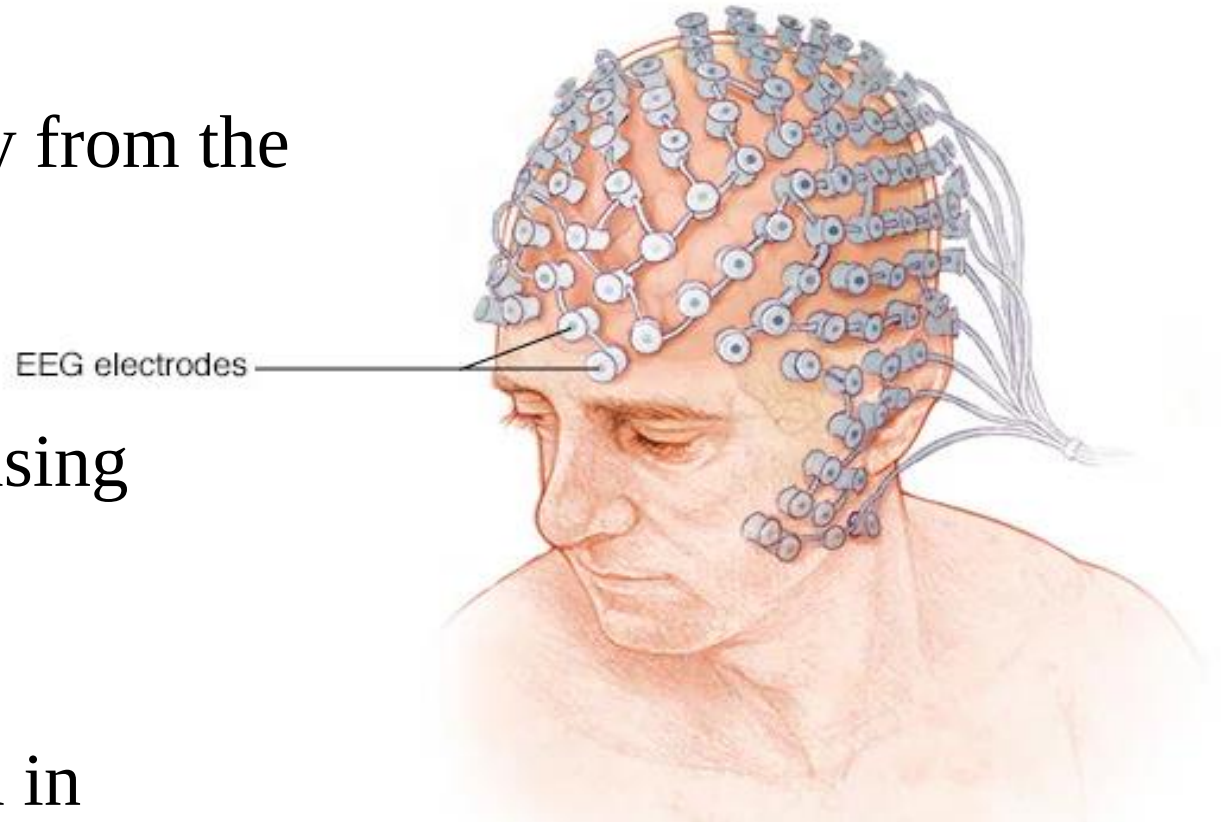
(a) Network Segregation



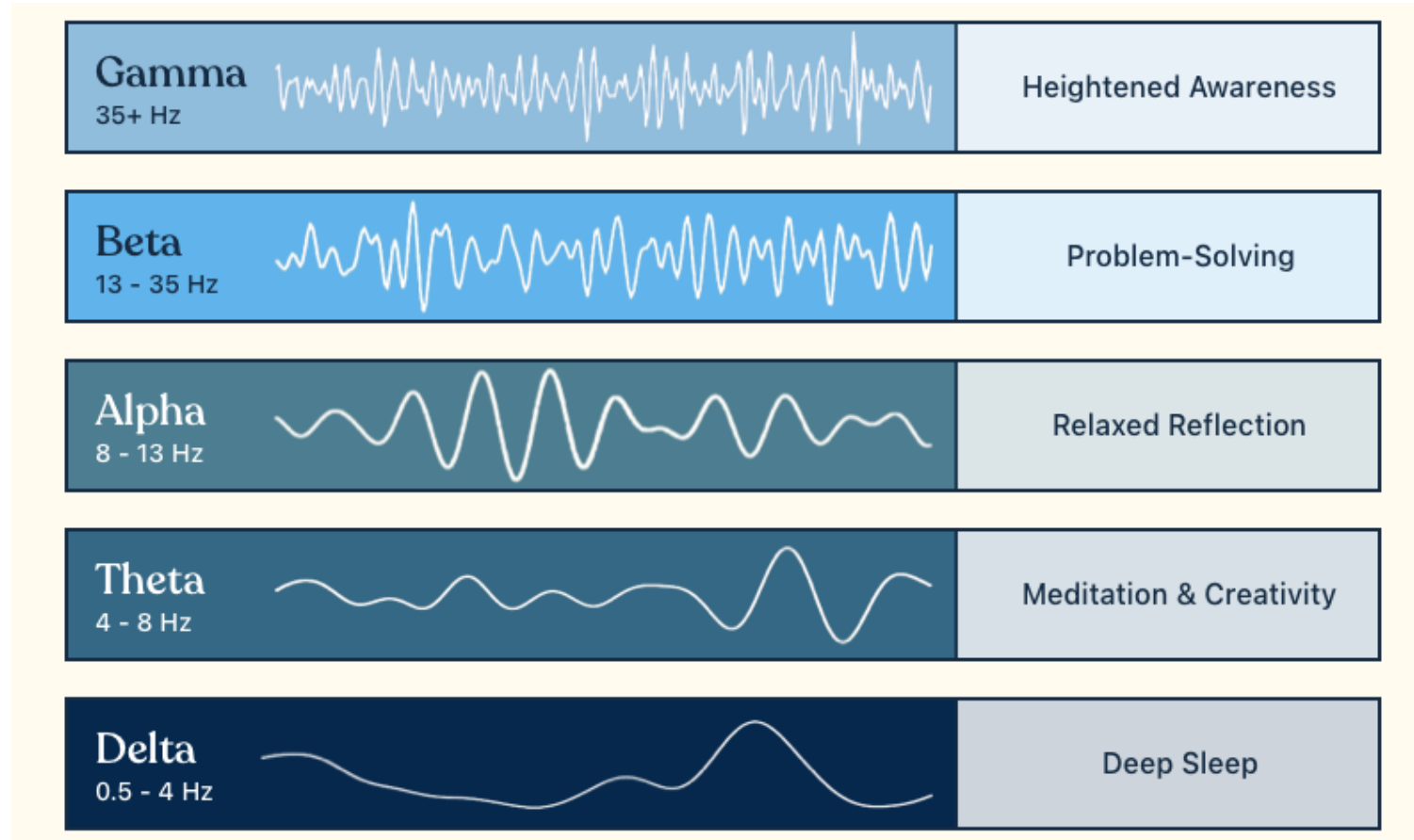
(b) Network Integration

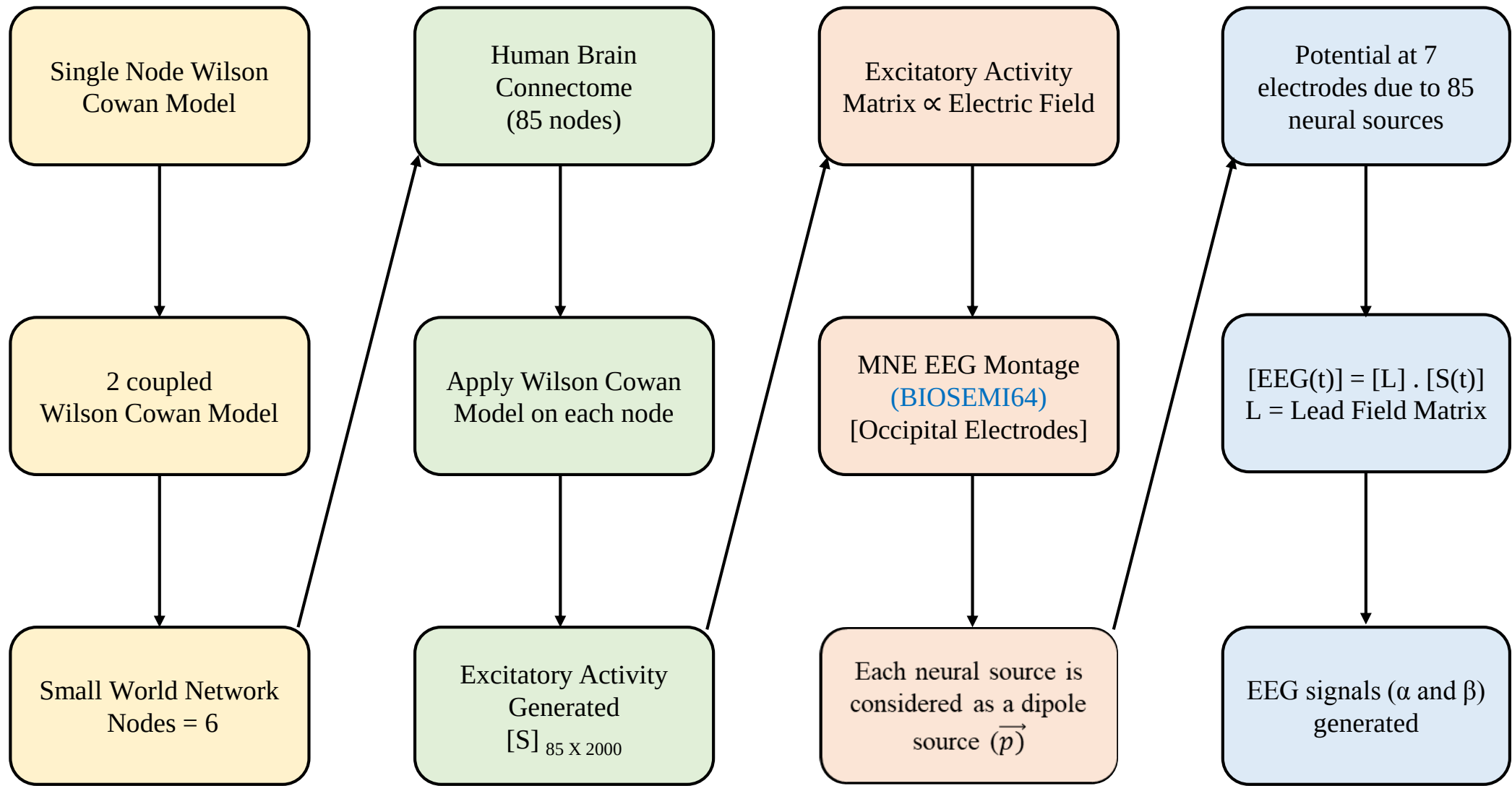
EEG: A Window into Brain Function

- EEG records electrical activity from the brain's surface.
- Measures neural oscillations using scalp electrodes.
- Non-invasive and widely used in research and clinical settings.



Brain waves with EEG





Wilson-Cowan Dynamics

$$\tau_u \dot{u}_i = -u_i + (\kappa_u - r_u u_i) s_u(u_i^{\text{in}})$$

$$\tau_v \dot{v}_i = -v_i + (\kappa_v - r_v v_i) s_v(v_i^{\text{in}})$$

$$u_i^{\text{in}} = c_{uu} u_i - c_{uv} v_i + \Sigma'(w_{ij}^{uu} u_j - w_{ij}^{uv} v_j) + I_u^{\text{ext}}$$

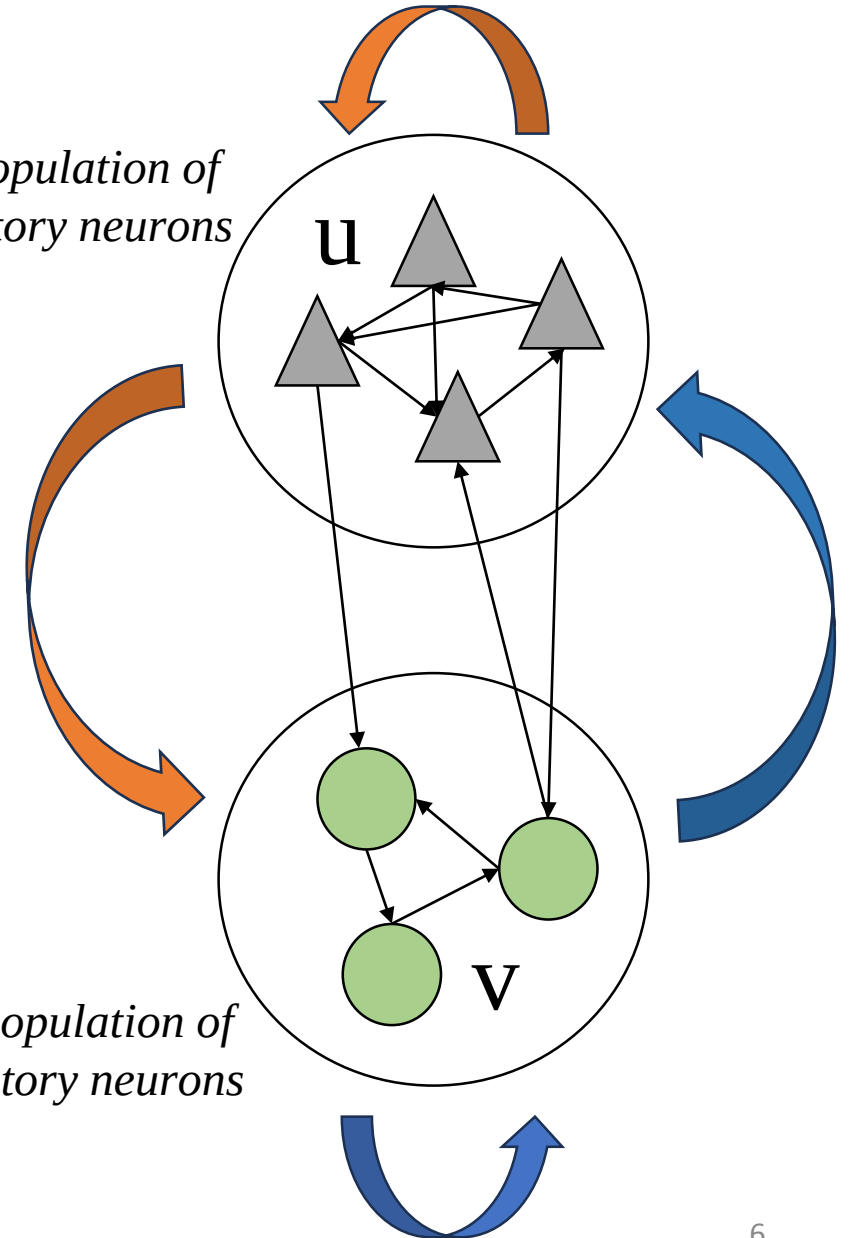
$$v_i^{\text{in}} = c_{vu} u_i - c_{vv} v_i + \Sigma'(w_{ij}^{vu} u_j - w_{ij}^{vv} v_j) + I_v^{\text{ext}}$$

Sigmoid Function:

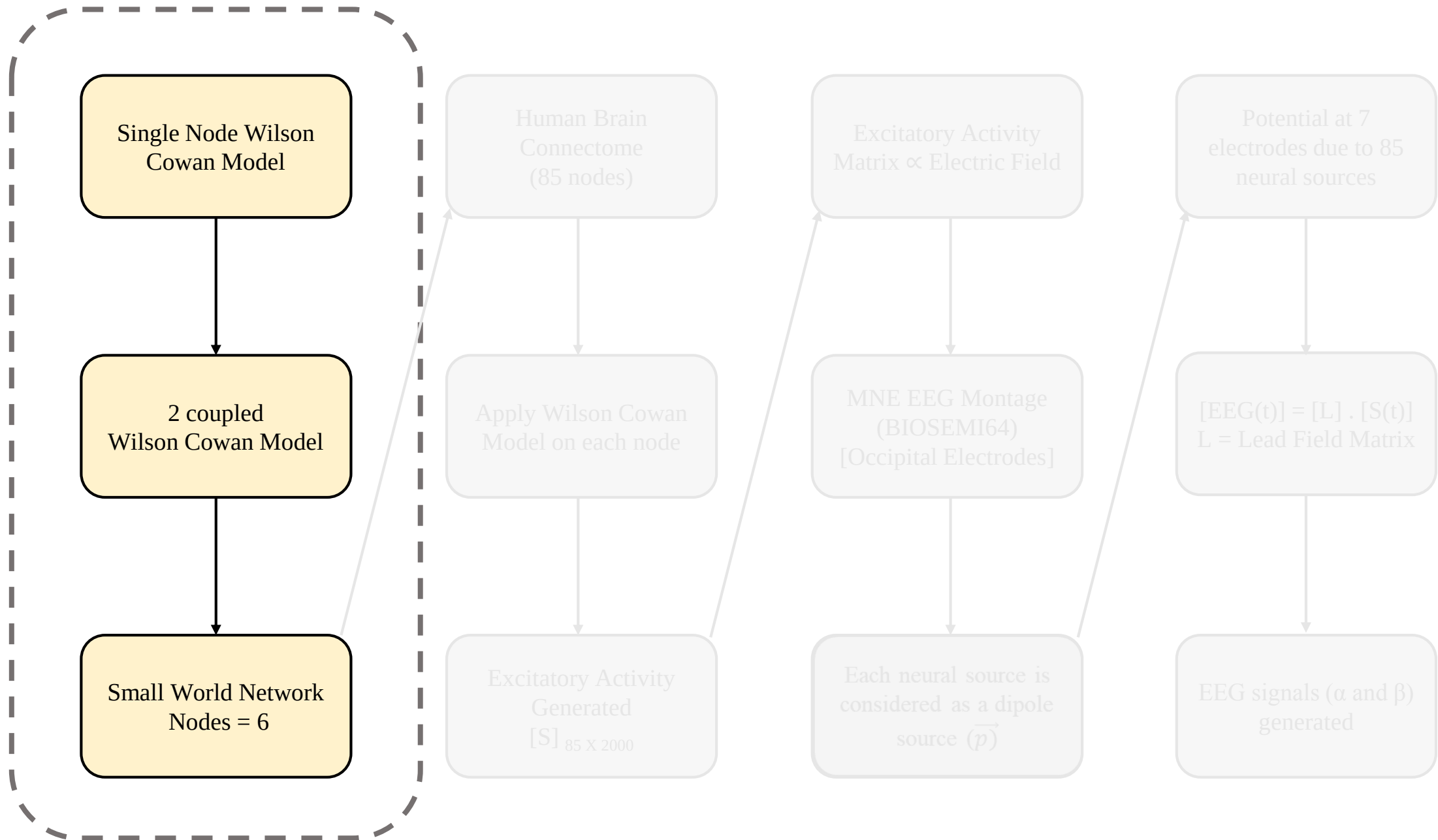
$$S_u(z) = [1 + \exp\{-a_\mu(z - \theta_\mu)\}]^{-1} + \kappa_\mu - 1$$

$$\kappa_\mu = 1 - [1 + \exp(a_\mu \theta_\mu)]^{-1}$$

*Sub-population of
excitatory neurons*



*Sub-population of
inhibitory neurons*

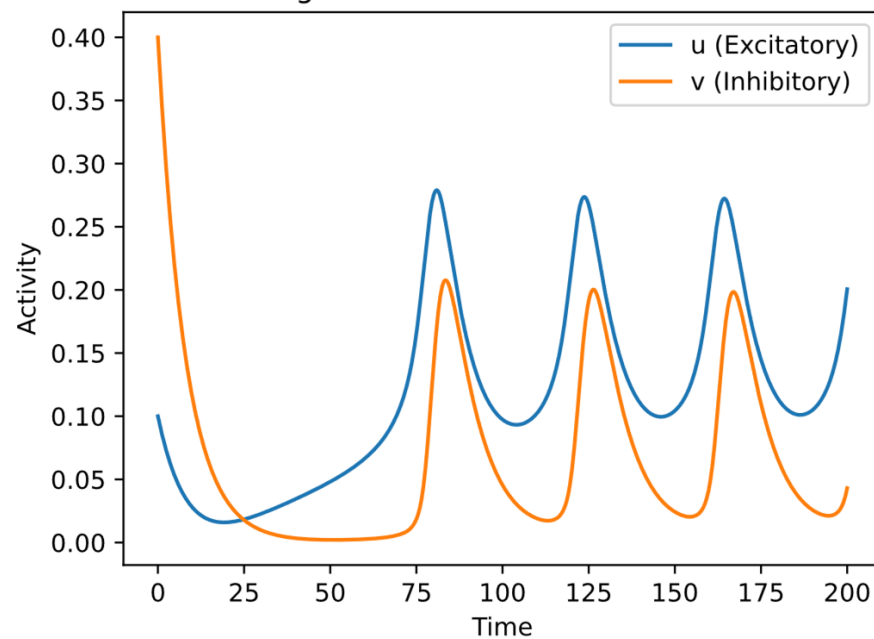


Single Node Wilson Cowan Model

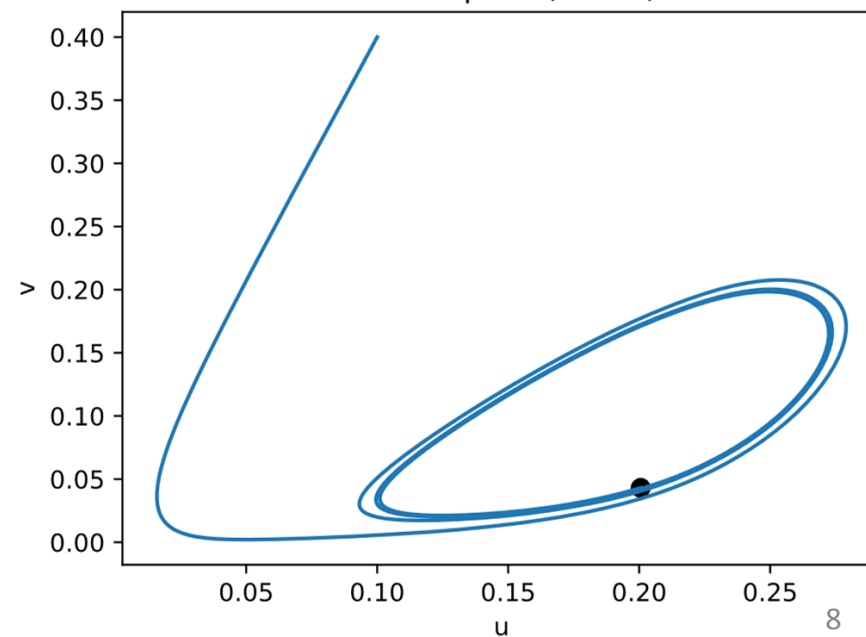
2 coupled
Wilson Cowan Model

Small World Network
Nodes = 6

Single WC Oscillator - Time Series



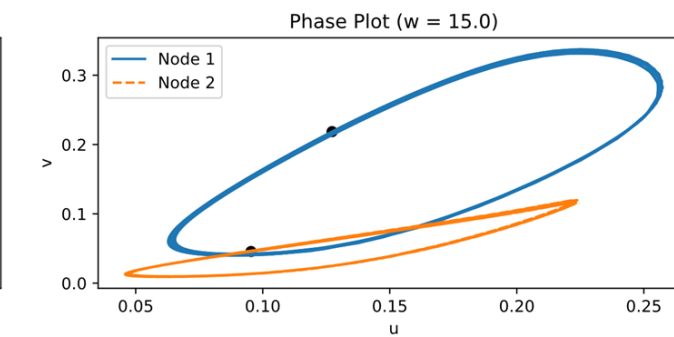
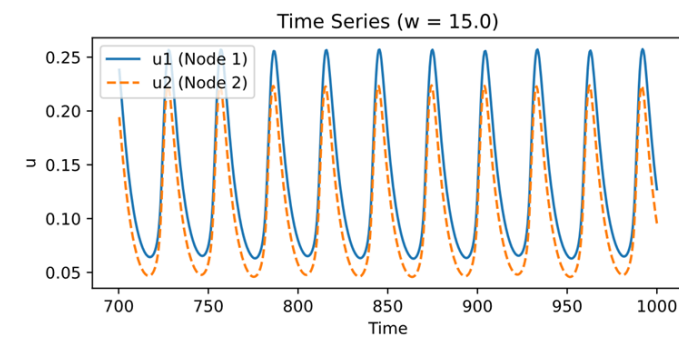
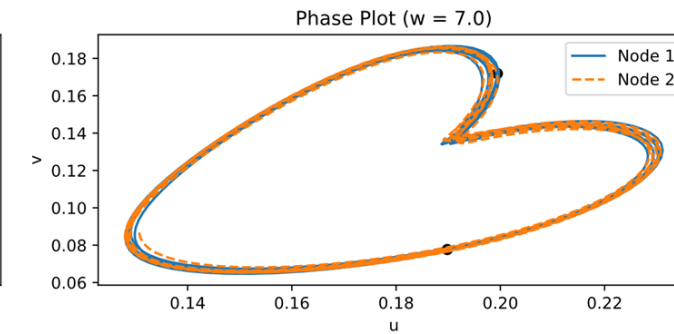
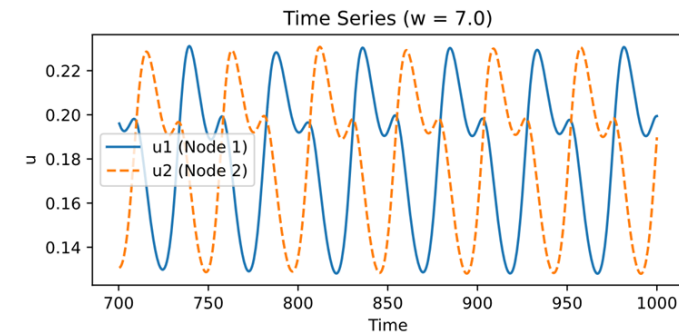
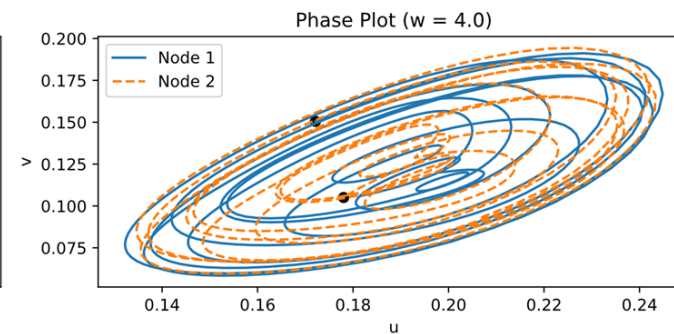
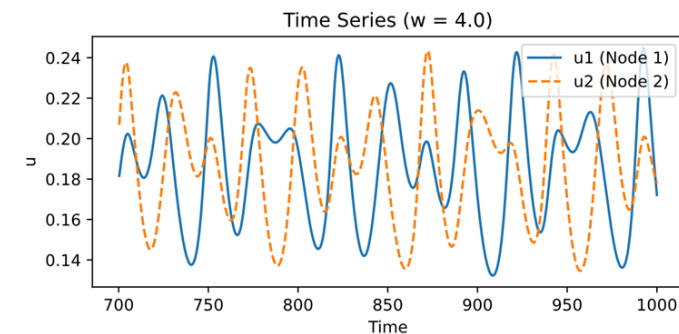
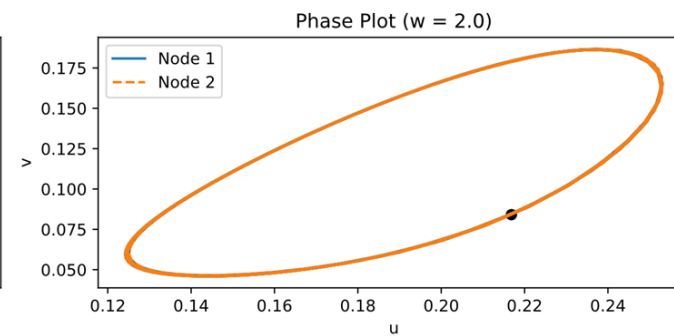
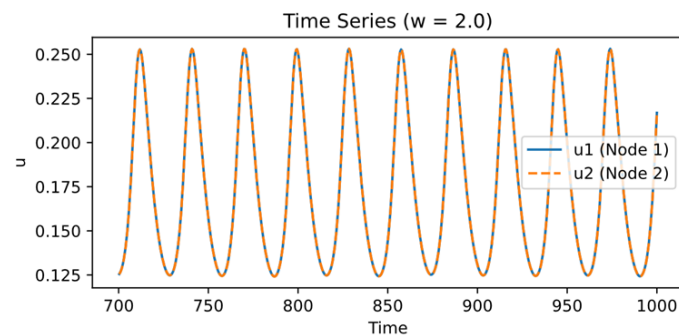
Phase Space (u vs v)

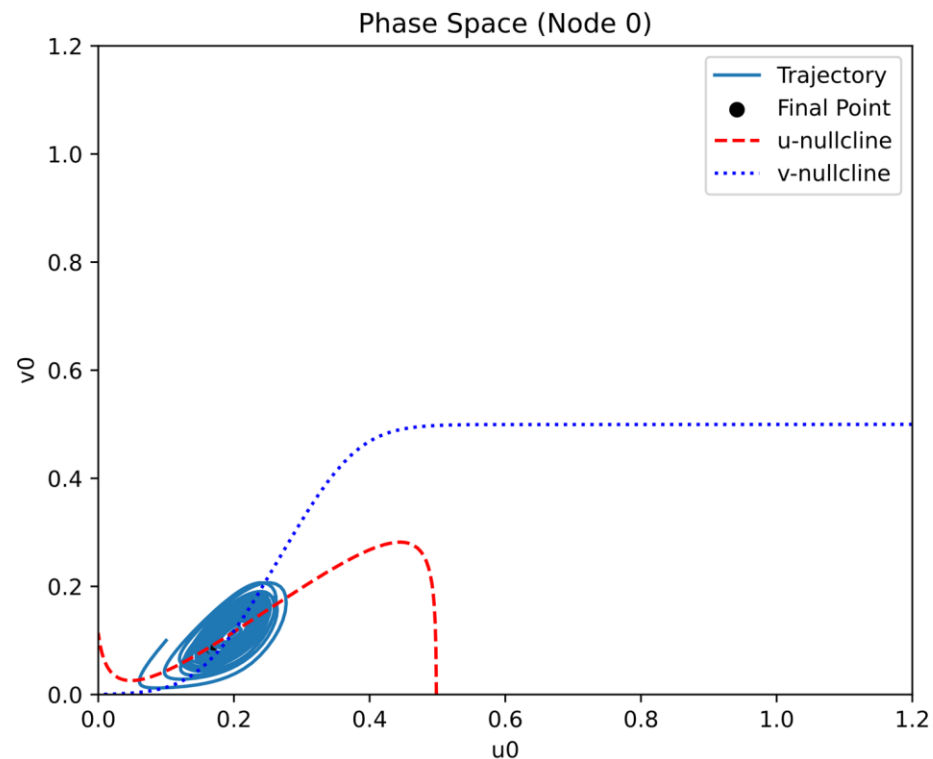
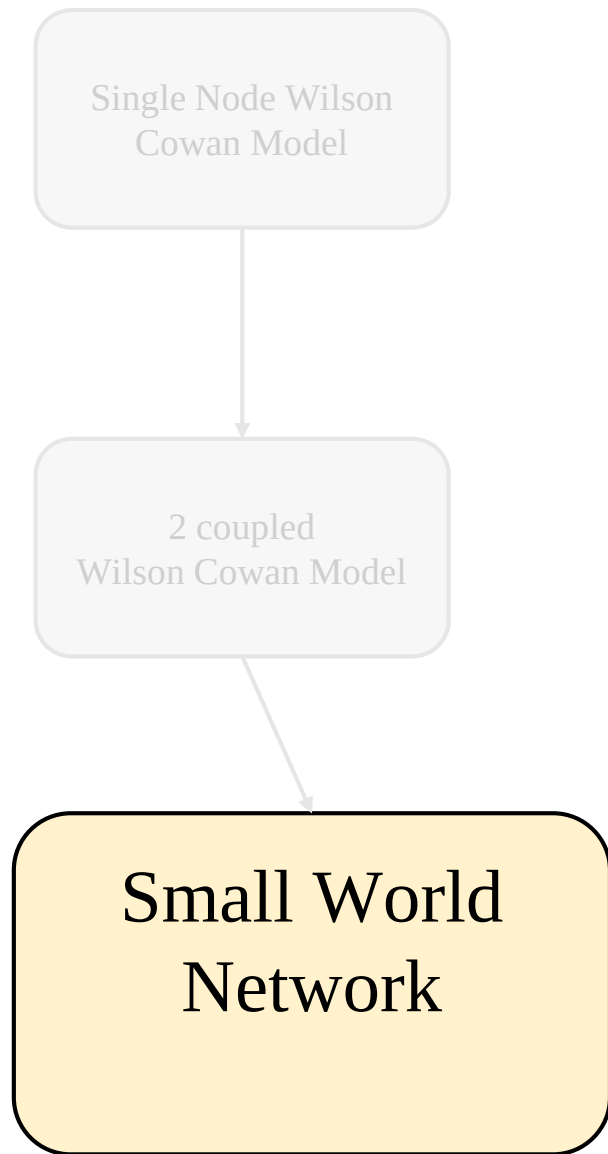


Single Node Wilson
Cowan Model

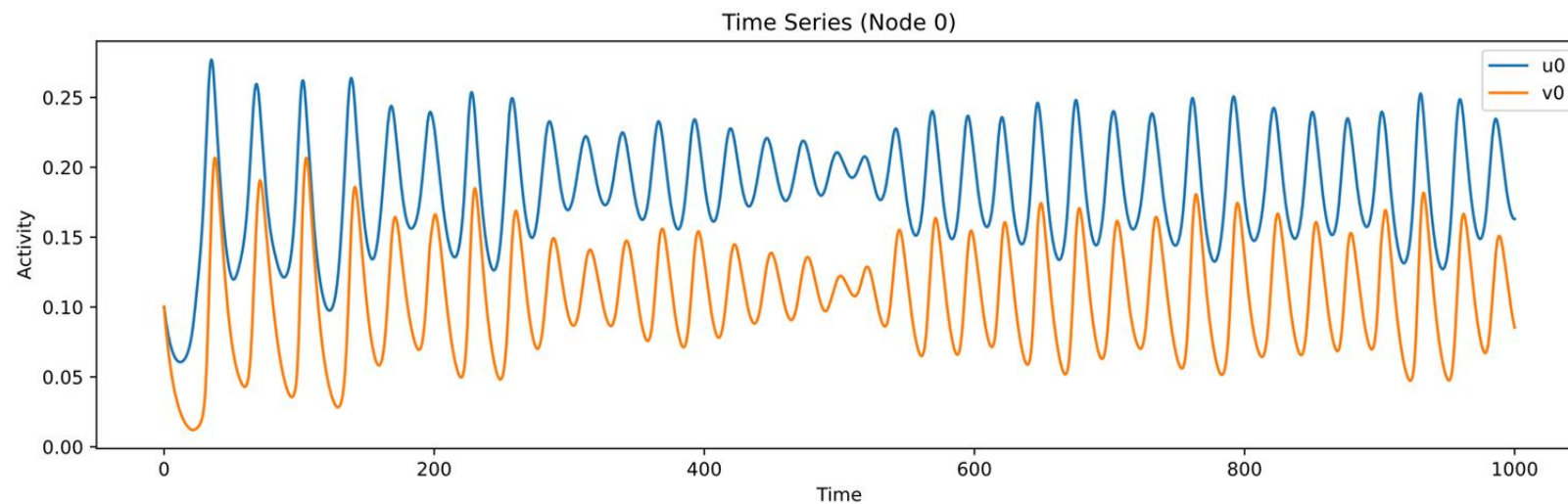
2 coupled
Wilson Cowan
Model

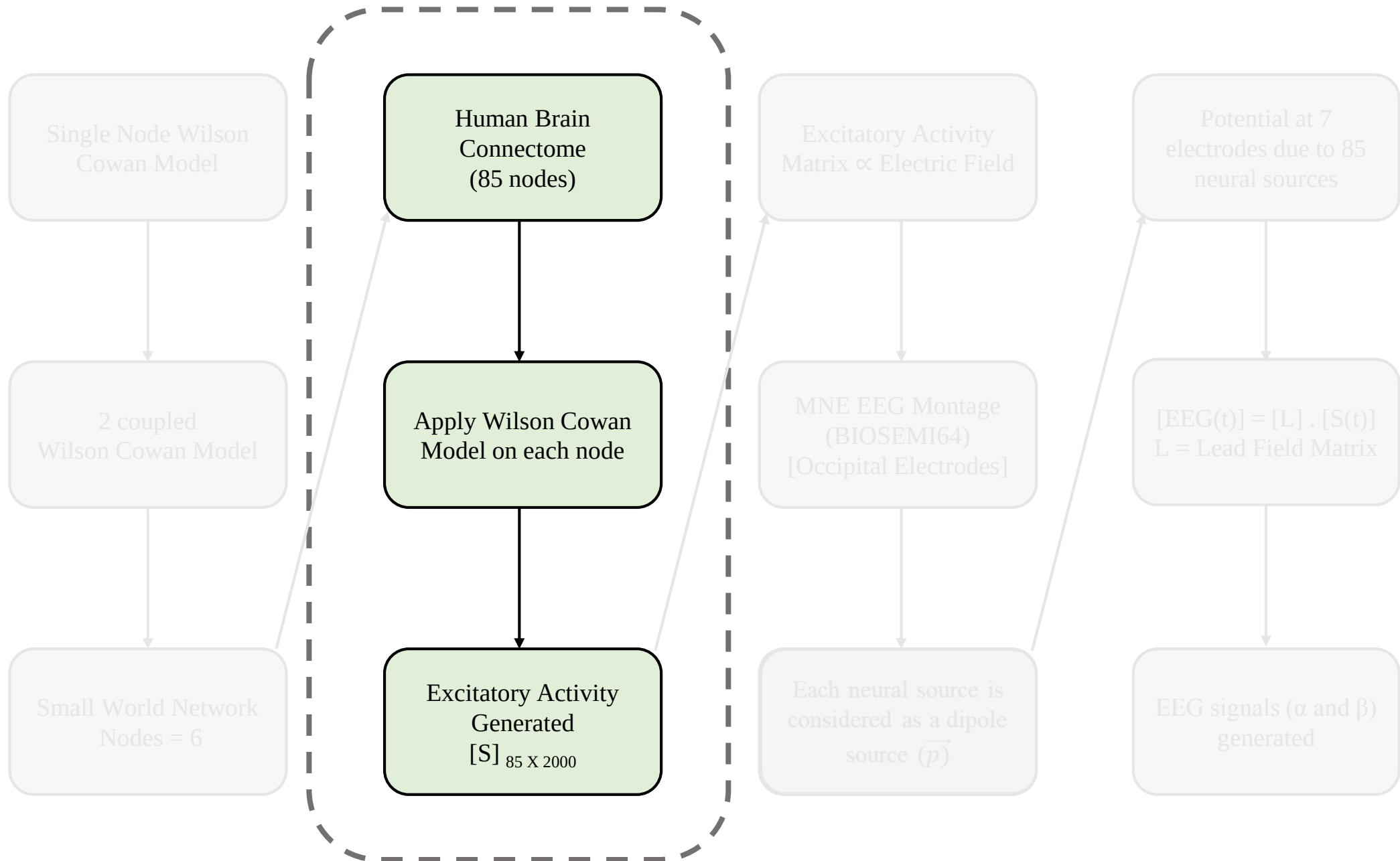
Small World Network
Nodes = 6





Phase Space and Time Series plot for Small World Network





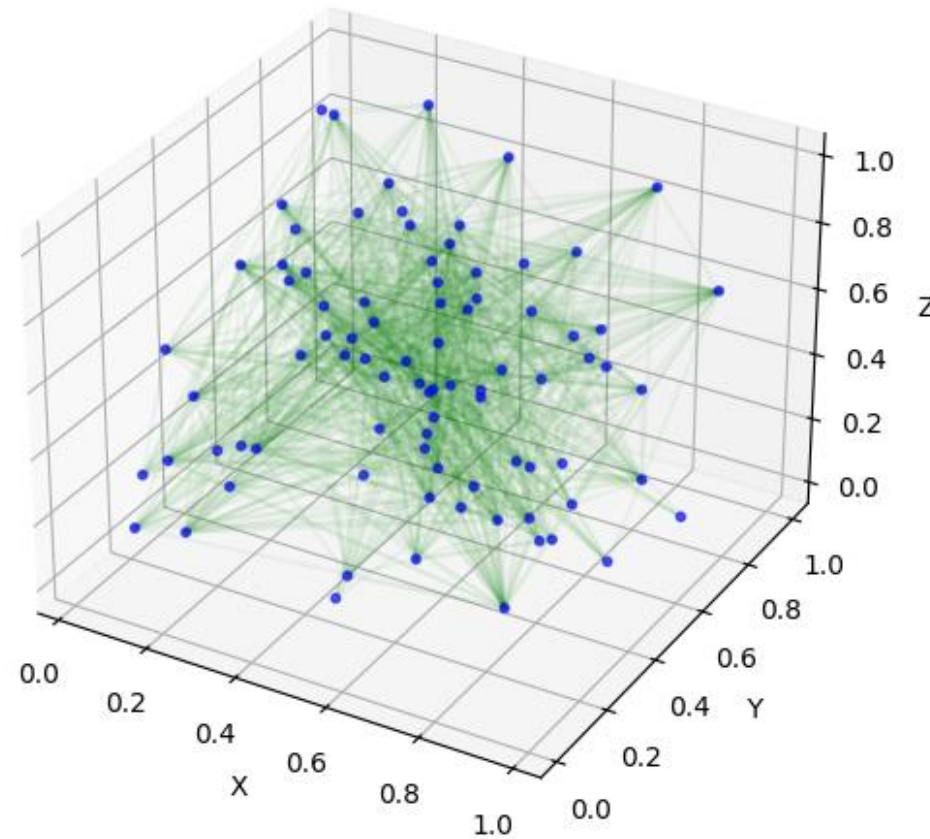
Spatial Layout of Nodes

Human Brain
Connectome
(85 nodes)

Apply Wilson Cowan
Model on each node

Excitatory Activity
Generated
[S] 85×2000

Visualisation of brain connectome in 3D space



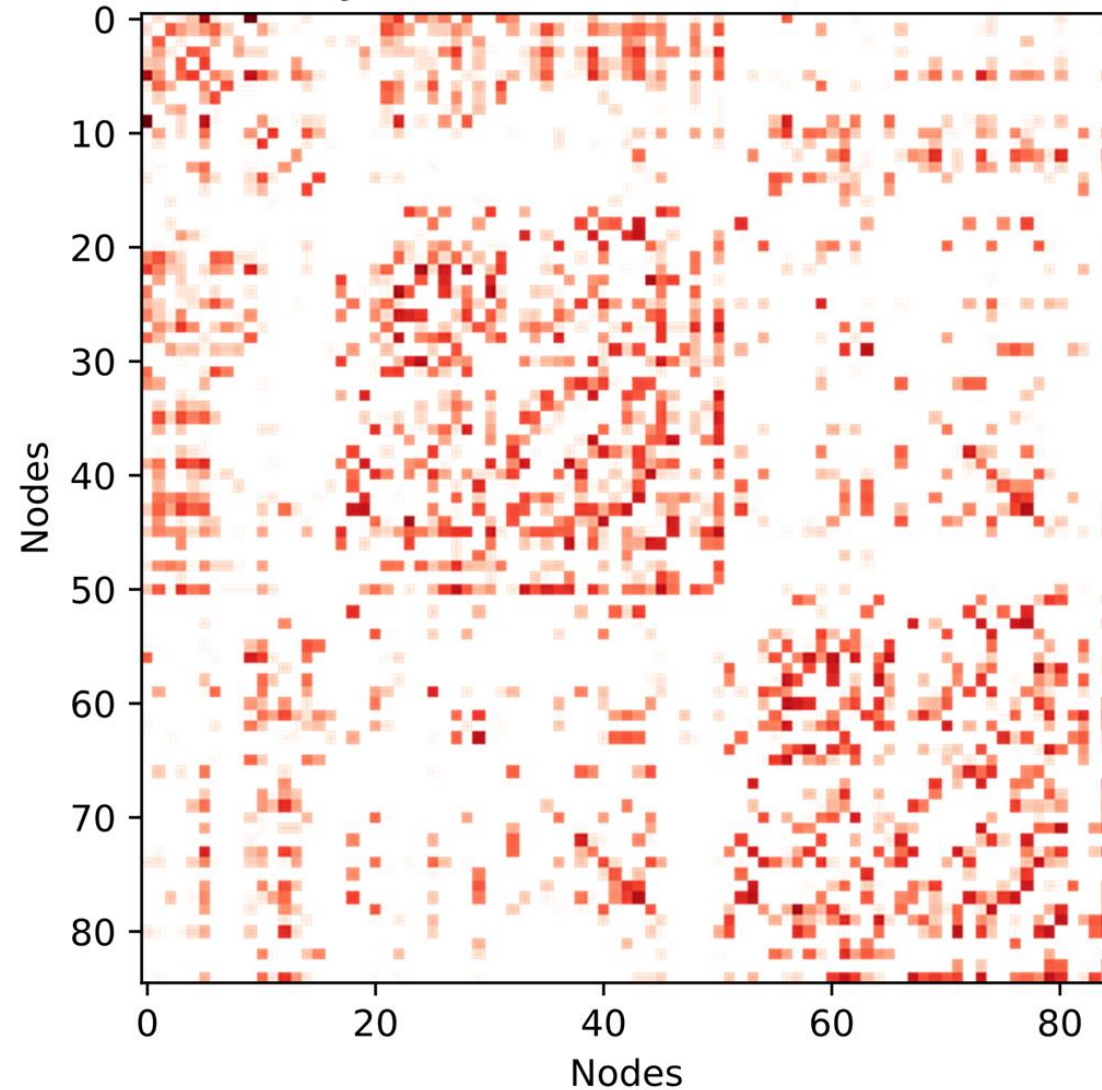
Data: <http://umcd.humanconnectomeproject.org/>

Connectivity Matrix of Human Brain Connectome

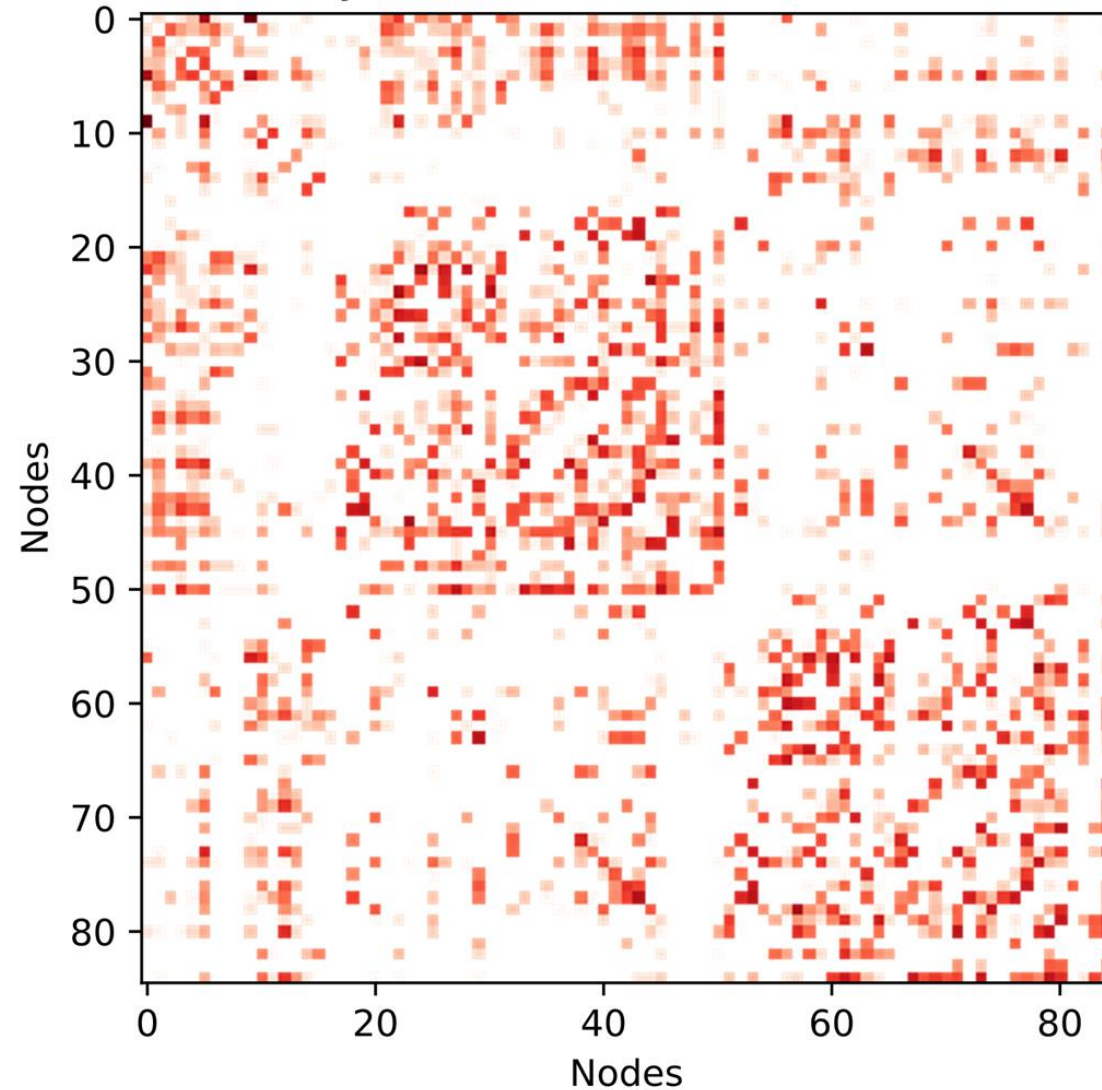
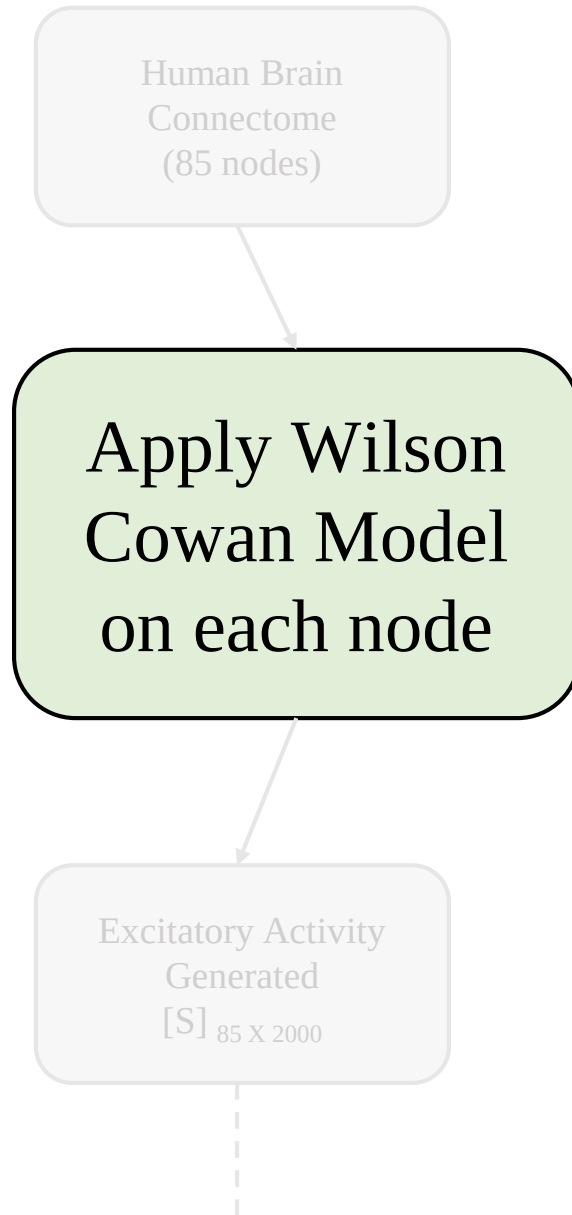
Human Brain
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Apply Wilson Cowan
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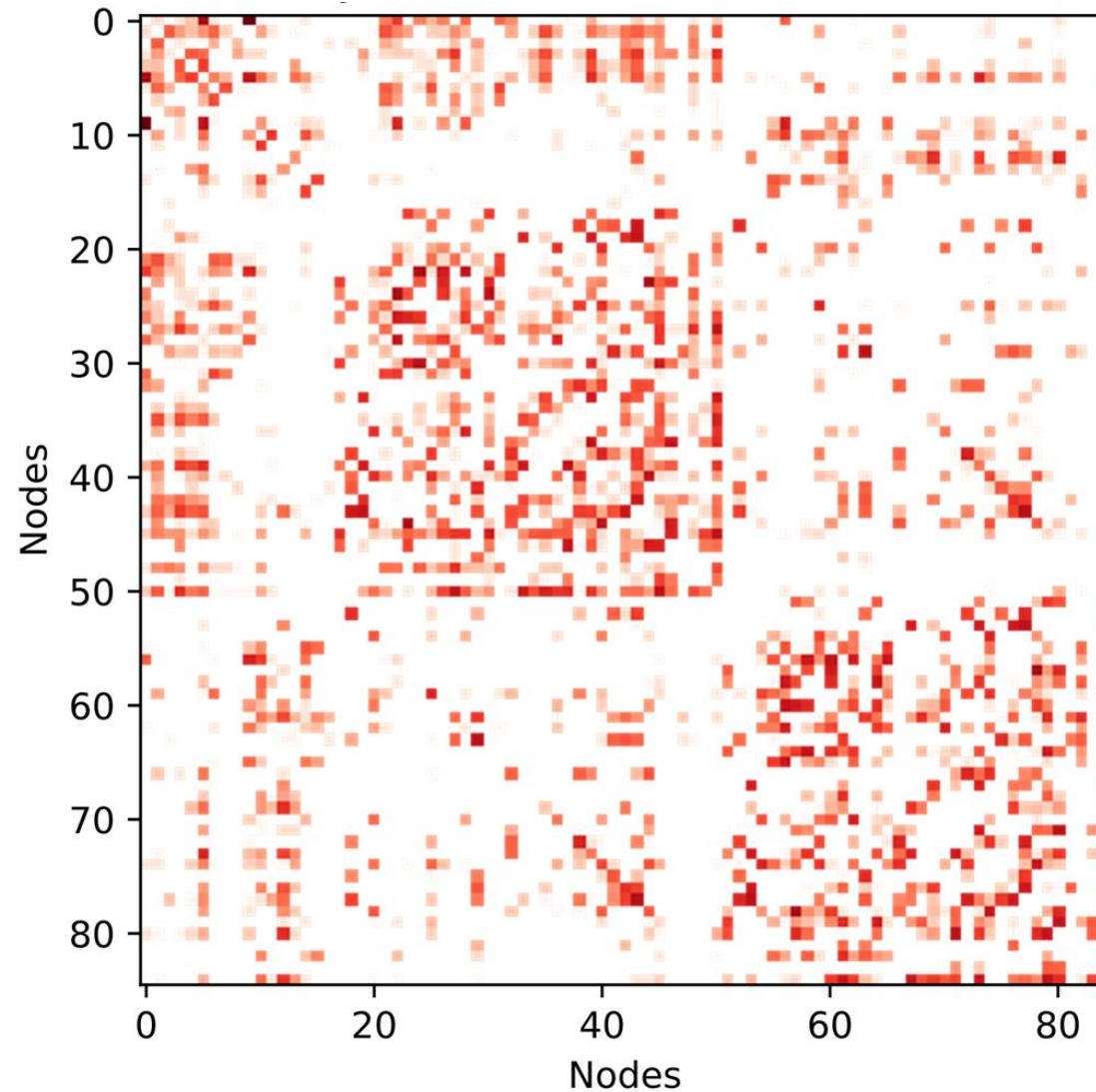
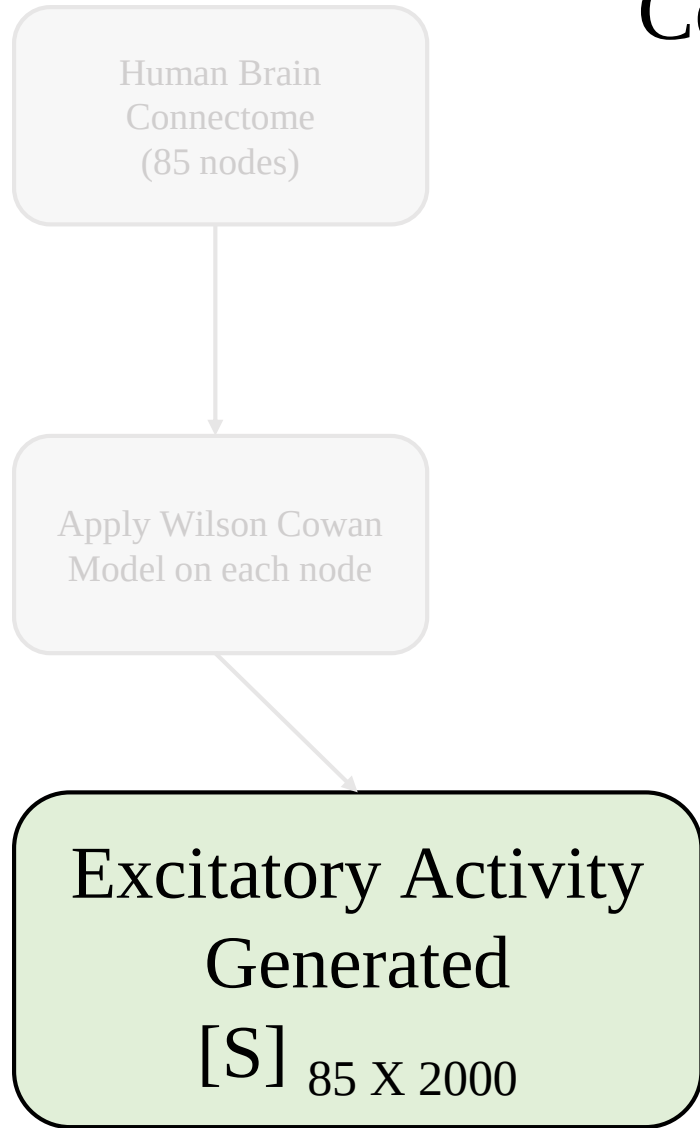
Excitatory Activity
Generated
[S] 85×2000

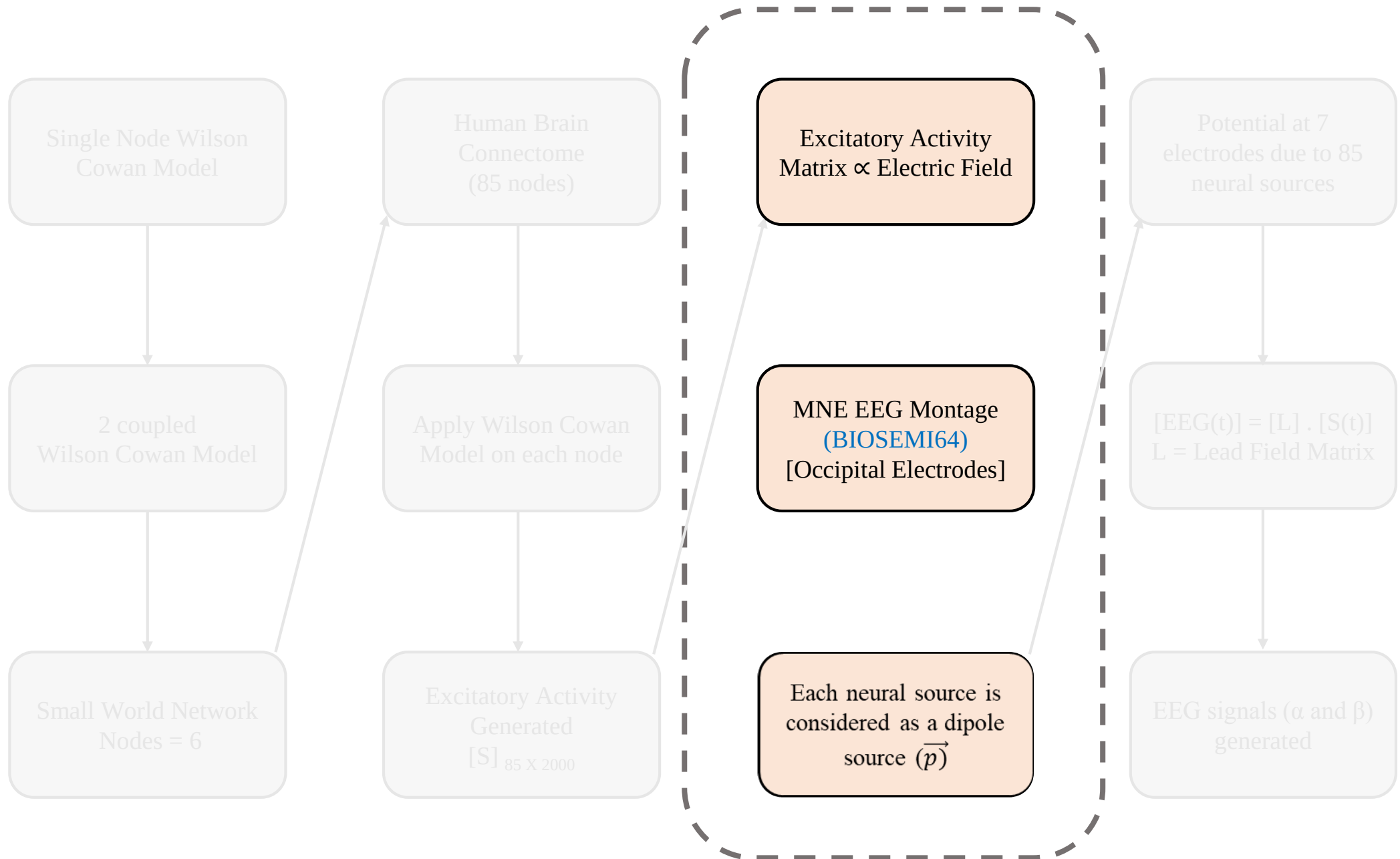


Connectivity Matrix of Human Brain Connectome



Connectivity Matrix of Human Brain Connectome



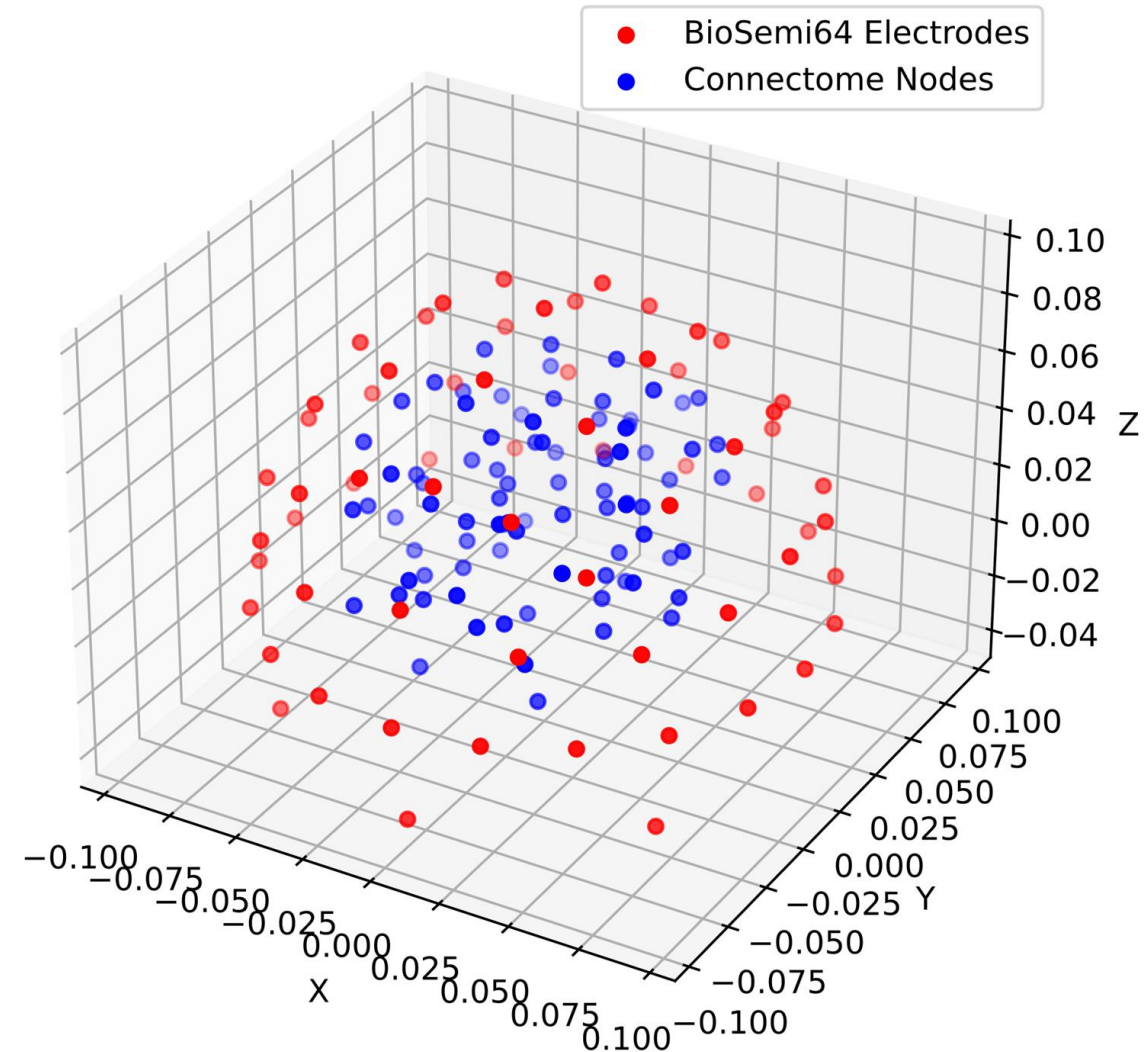


Excitatory
Activity Matrix
 $\propto$ Electric Field

MNE EEG Montage
(BIOSEMI64)
[Occipital Electrodes]

Each neural source is
considered as a dipole
source $(\vec{p})$

MNE EEG Montage: BioSemi64

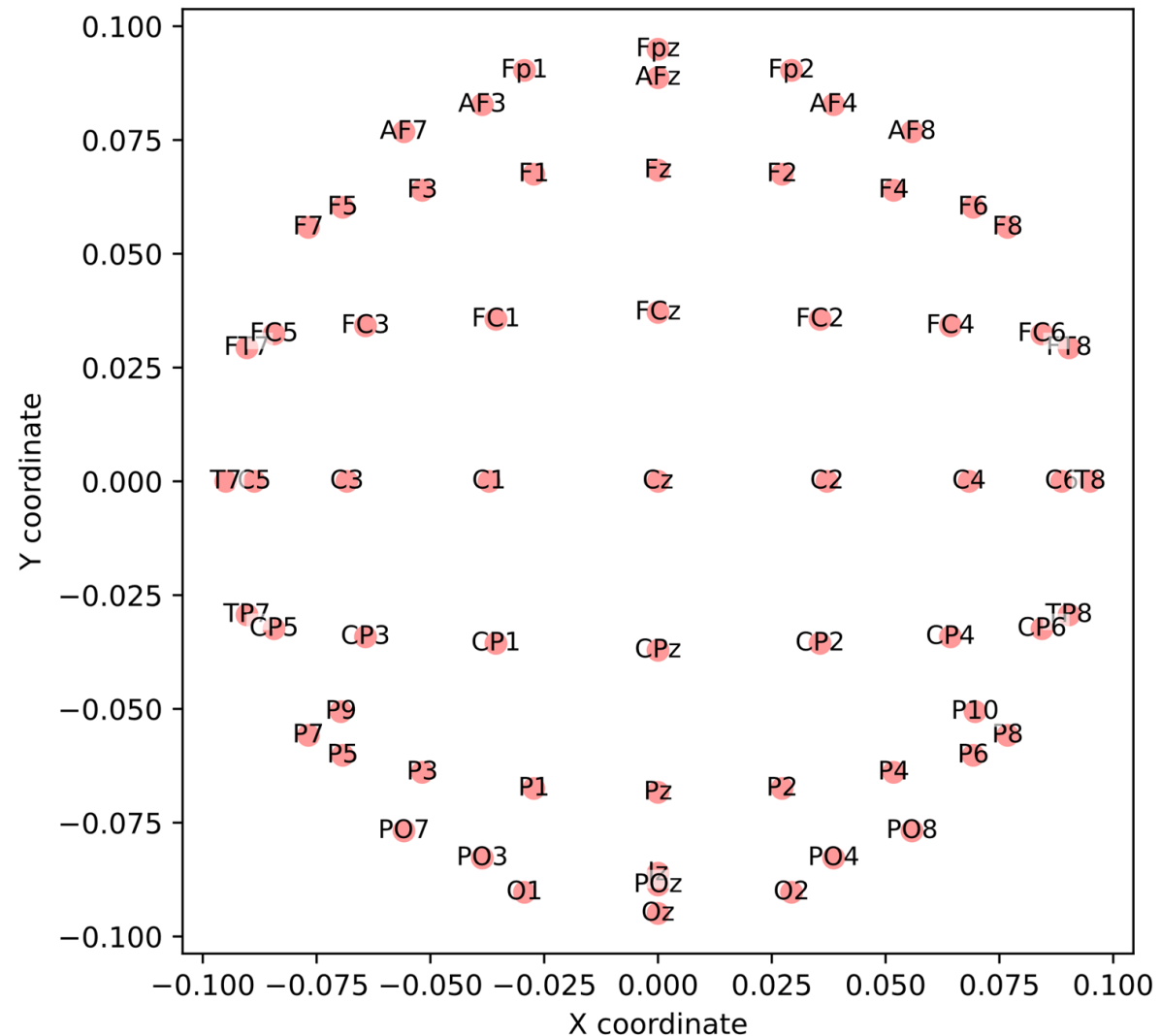


MNE EEG Montage: BioSemi64

Excitatory Activity
Matrix $\propto$ Electric Field

MNE EEG Montage
(BIOSEMI64)
[Occipital
Electrodes]

Each neural source is
considered as a dipole
source ($\vec{p}$)

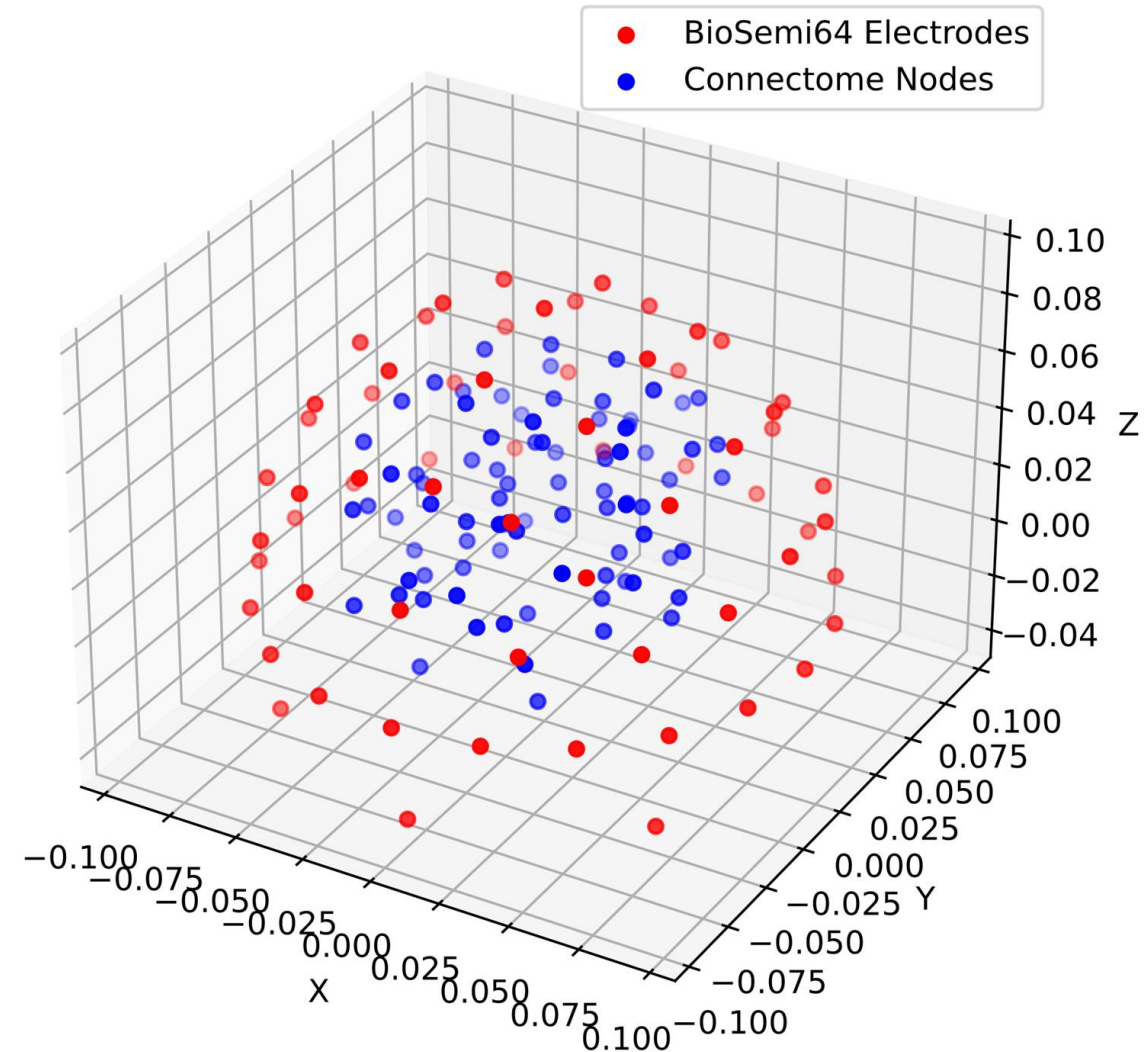


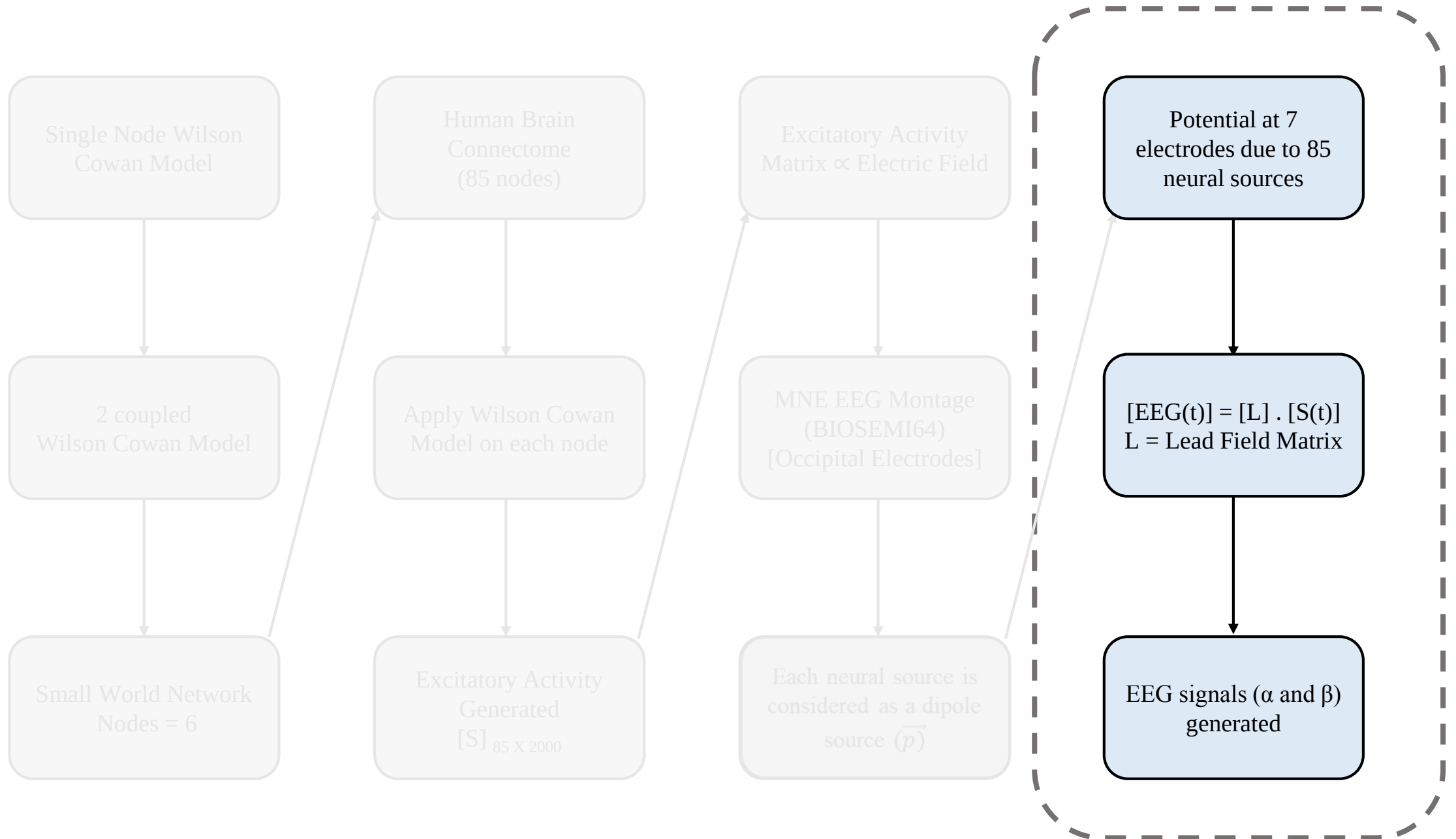
MNE EEG Montage: BioSemi64

Excitatory Activity
Matrix $\propto$ Electric Field

MNE EEG Montage
(BIOSEMI64)
[Occipital Electrodes]

Each neural source
is considered as a
dipole source ($\vec{p}$)

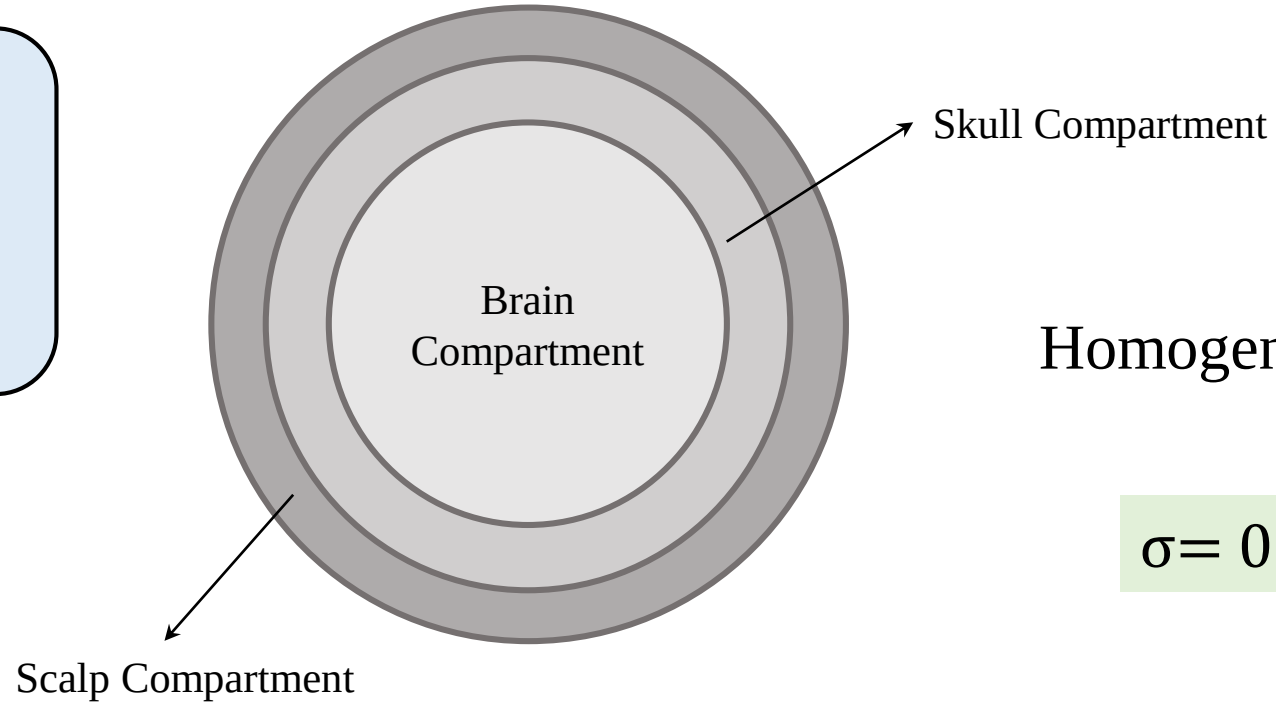




Potential at 7
electrodes due to
85 neural sources

L = Lead Field Matrix

EEG signals (α and β)
generated



Homogeneous Model

$$\sigma = 0.33 \text{ S/m}$$

$$V(\vec{r}_e) = \frac{1}{4\pi\sigma} \frac{\vec{p} \cdot (\vec{r}_e - \vec{r}_d)}{|\vec{r}_e - \vec{r}_d|^3}$$

$\vec{p}$: Dipole moment (in A.m)

$\vec{r}_e$: Position of the electrode (in mm)

$\vec{r}_d$: Position of the dipole (neural source)
(in mm)

Potential at 7
electrodes due to 85
neural sources

Lead Field Matrix linearly maps the neural source activity to the measured potentials at the electrodes.

L = Lead Field Matrix

Each column of L_{ij} gives us the potential pattern at electrode i due to unit dipole in direction j .

$$\vec{p} = [0, 0, 1] \text{ (in z-direction)}$$

EEG signals (α and β)
generated

$$[\text{EEG}(t)]_{7 \times 2000} = [L]_{7 \times 85} \cdot [S(t)]_{85 \times 2000}$$

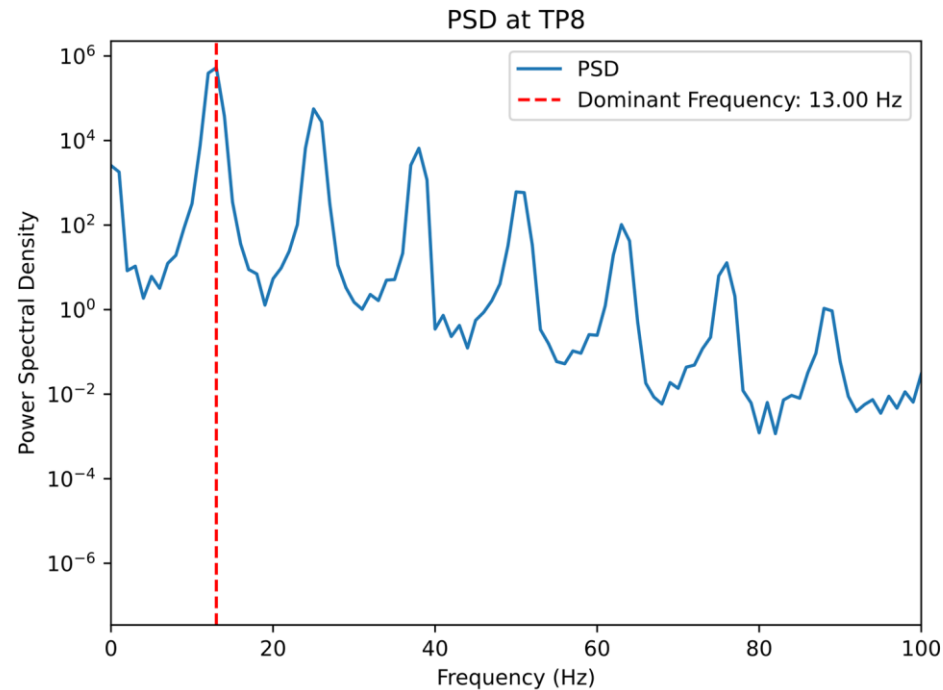
Power Spectral Density and Excitatory Activity

Potential at 7 electrodes due to 85 neural sources

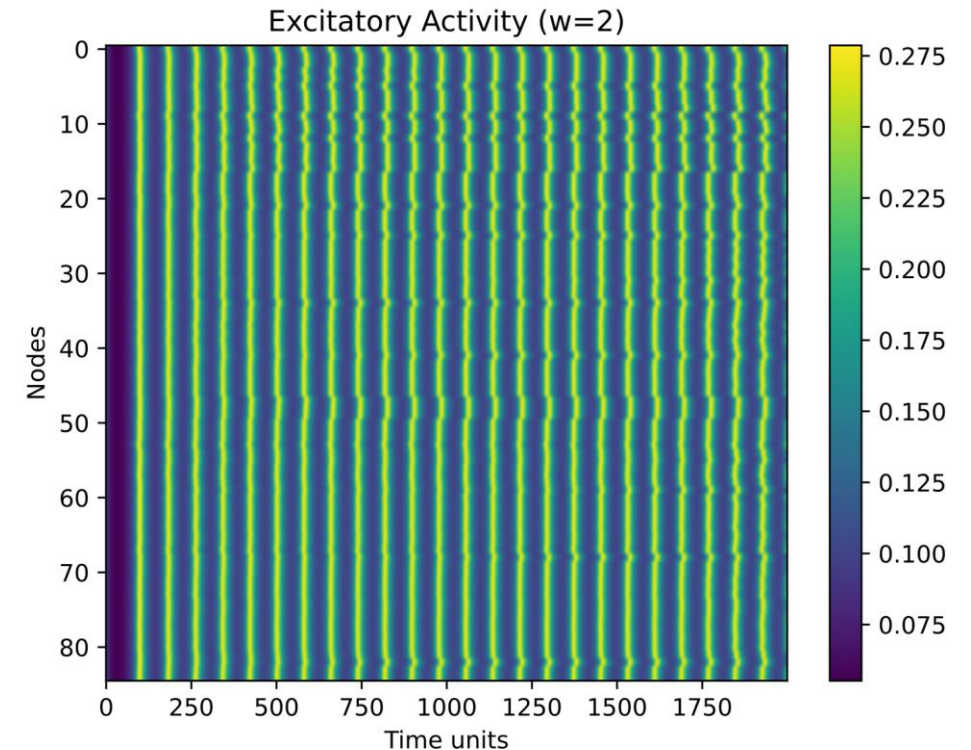
$$[\text{EEG}(t)] = [L] \cdot [S(t)]$$

L = Lead Field Matrix

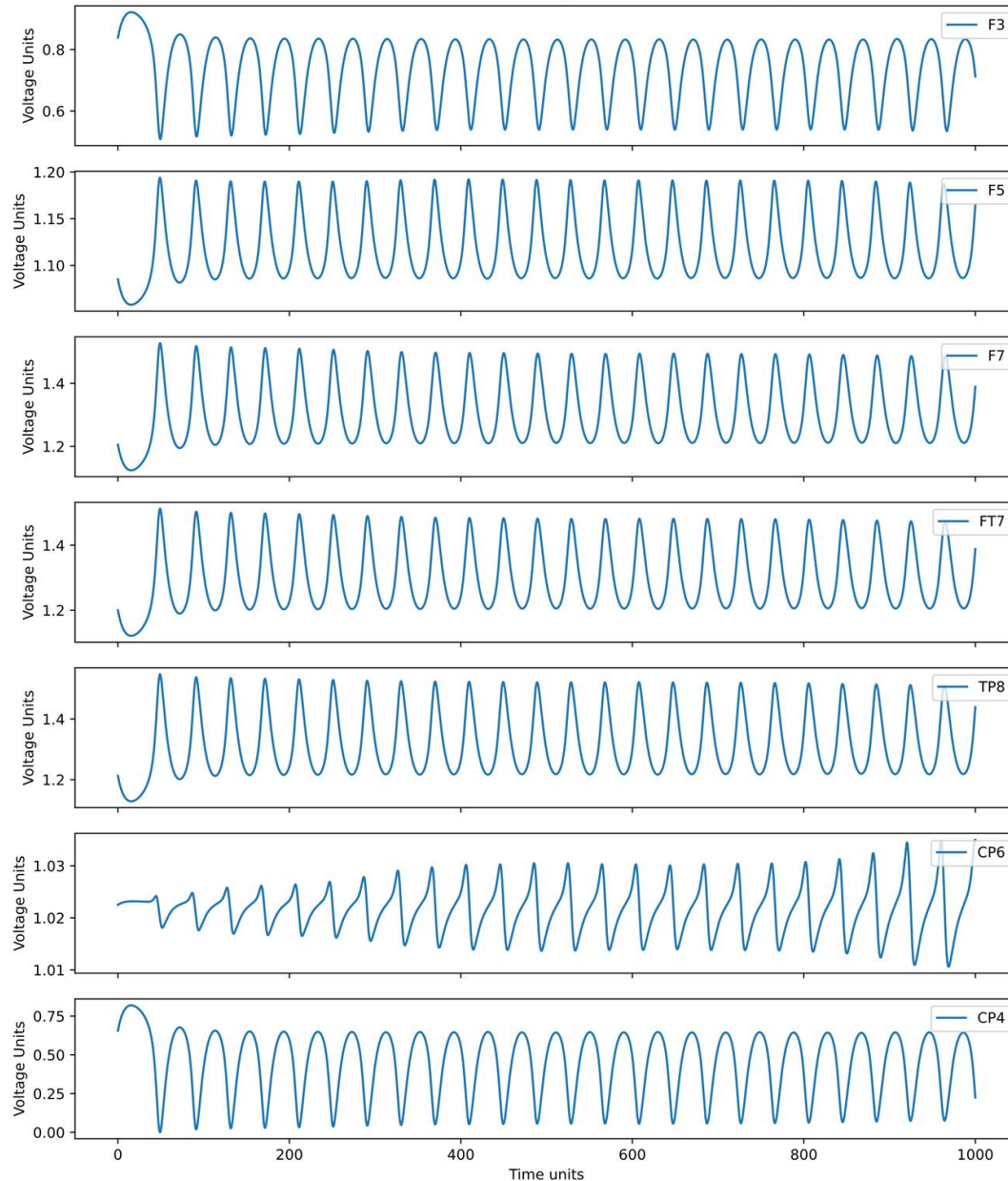
EEG signals (α and β) generated



Coupling Strength (w) = 2

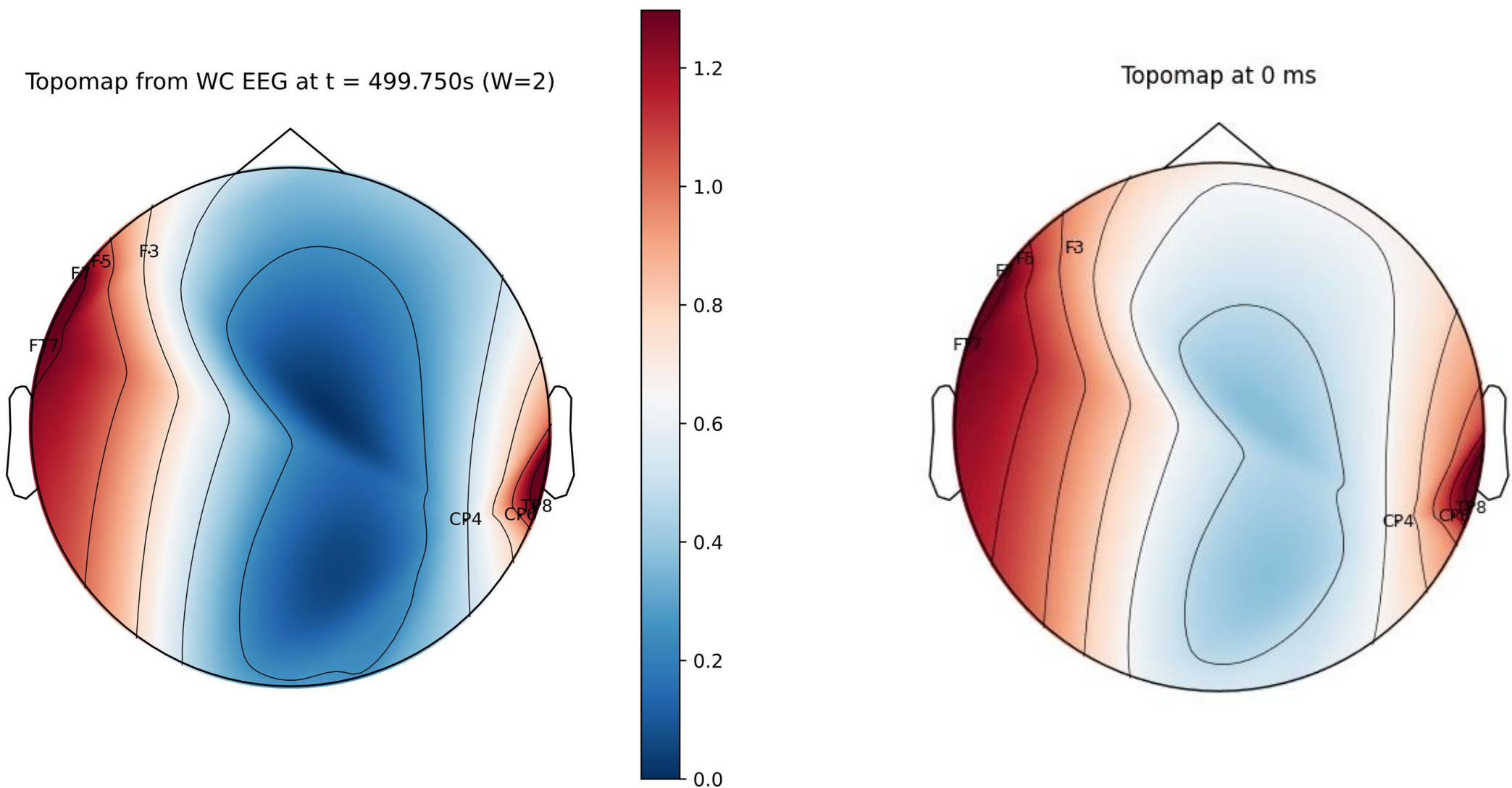


Simulated EEG from WC model (7 random electrodes)



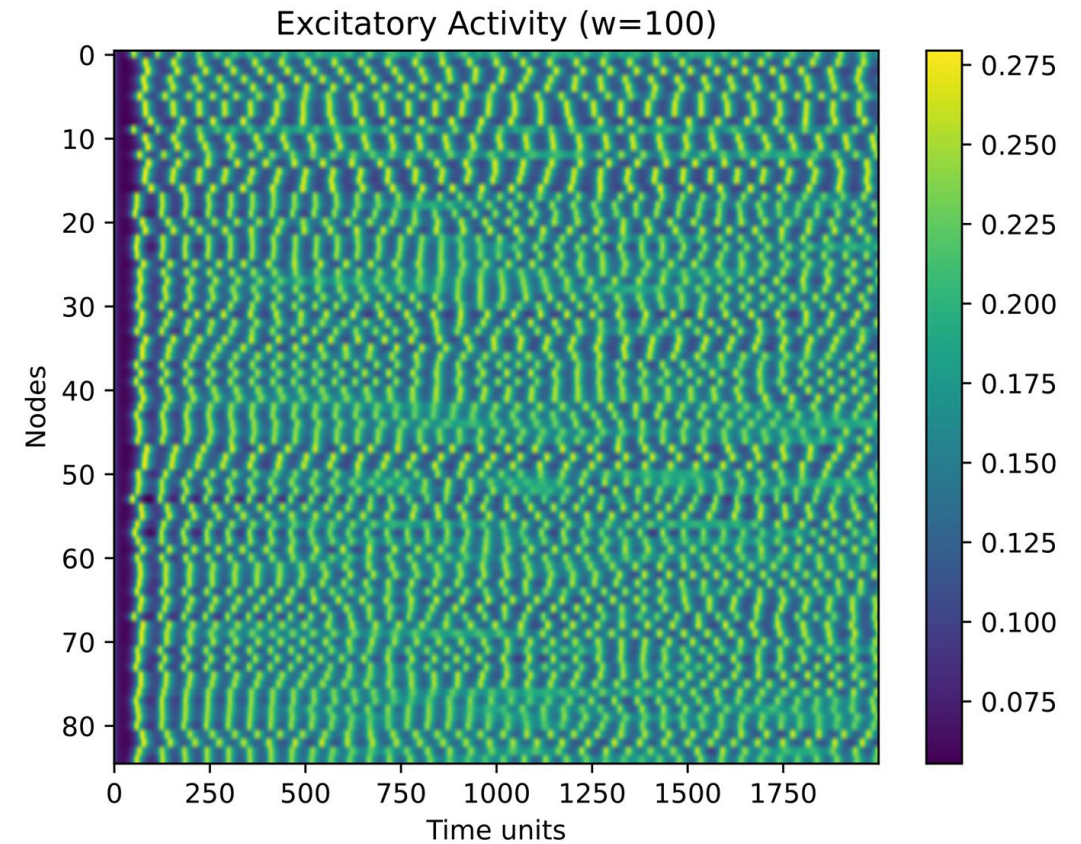
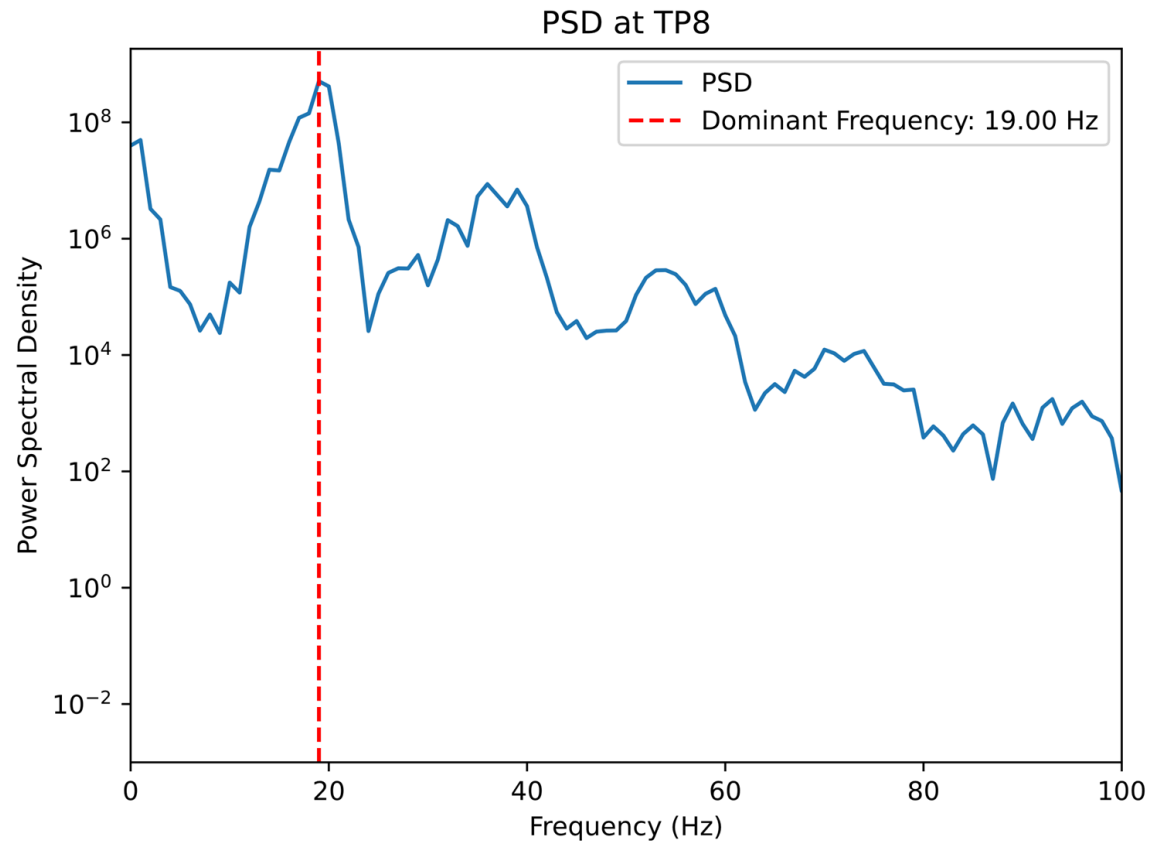
Coupling Strength (w) = 2

EEG Topomap from Wilson-Cowan Model

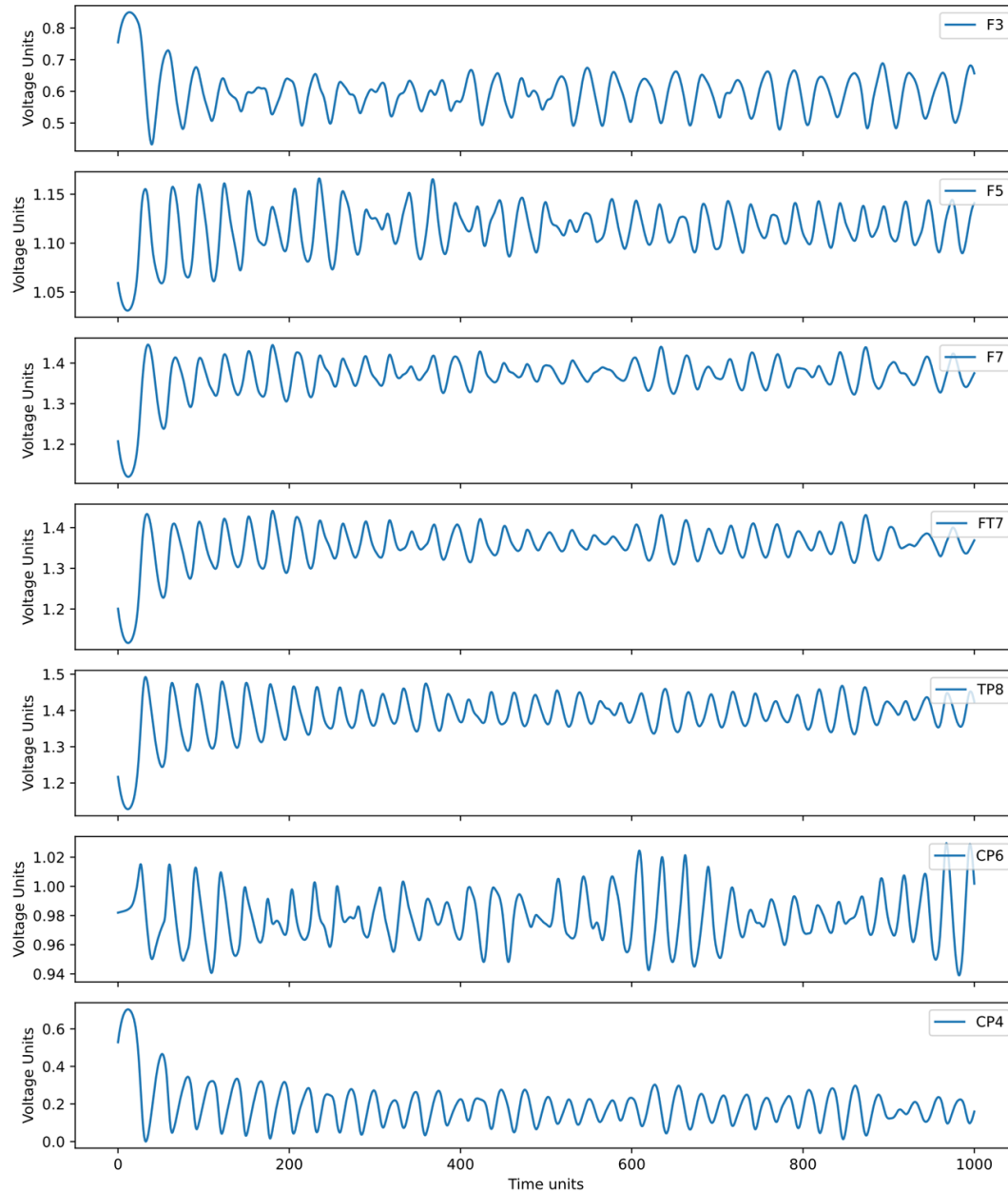


Power Spectral Density and Excitatory Activity

Coupling Strength (w) = 100

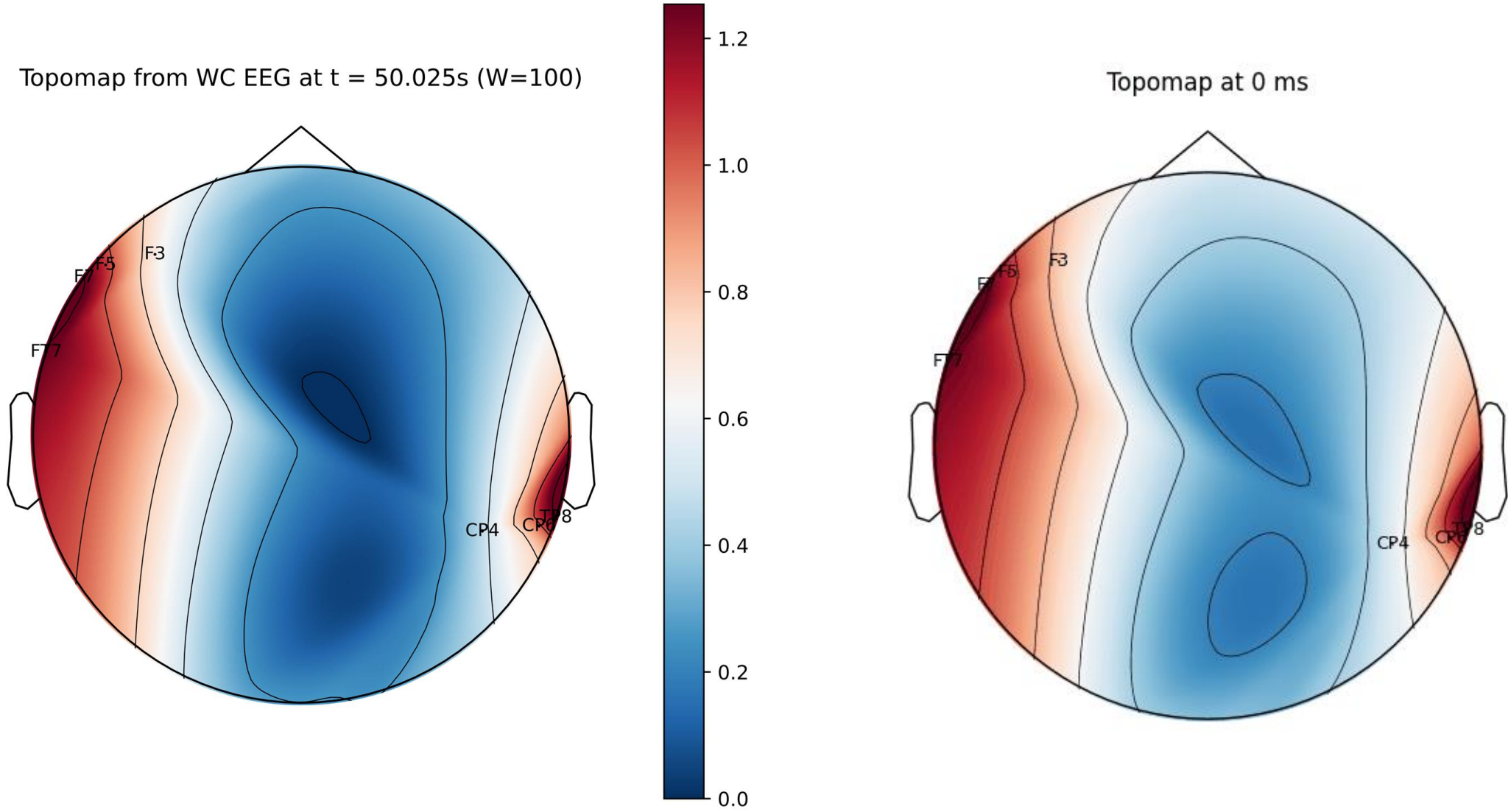


Simulated EEG from WC model (7 random electrodes)

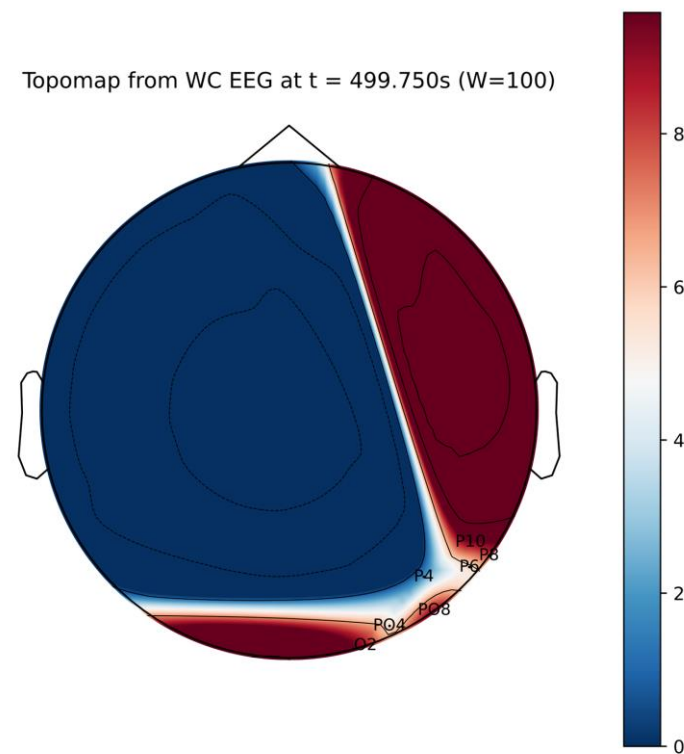
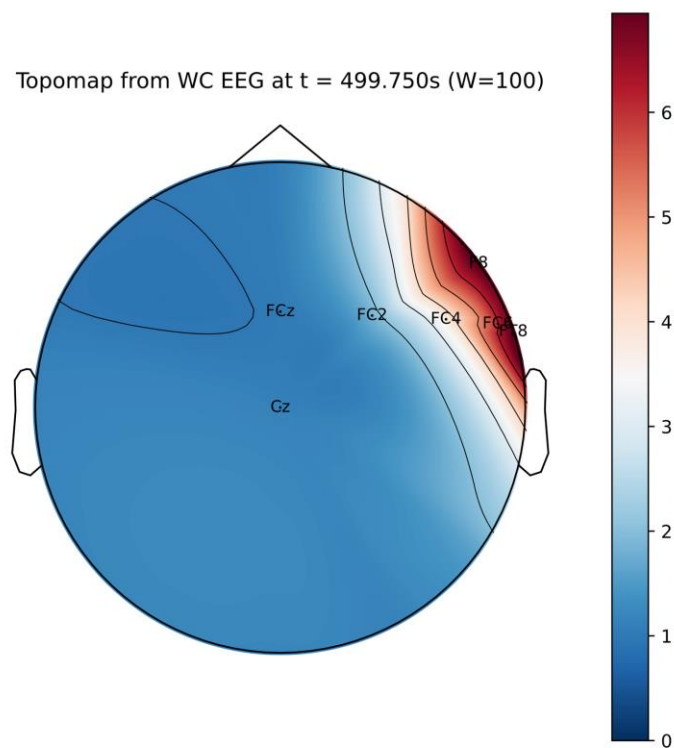
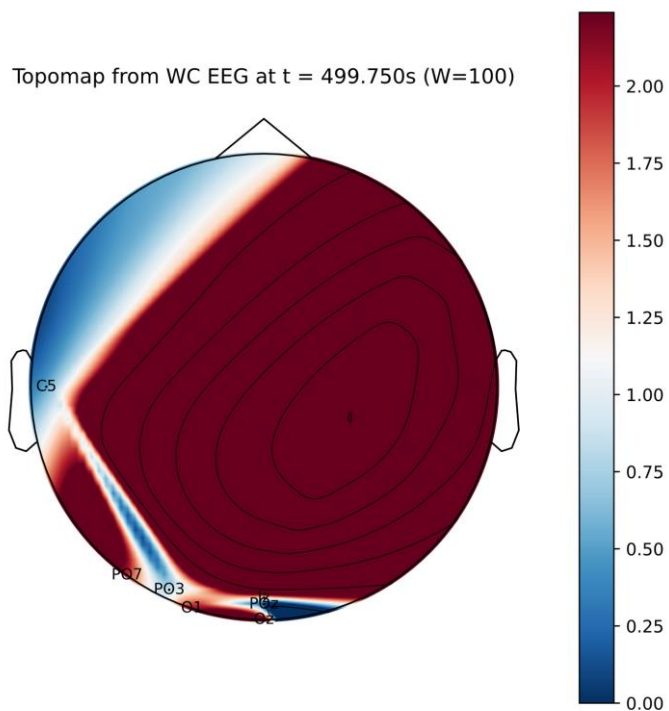


Coupling Strength (w) = 100

EEG Topomap from Wilson-Cowan Model

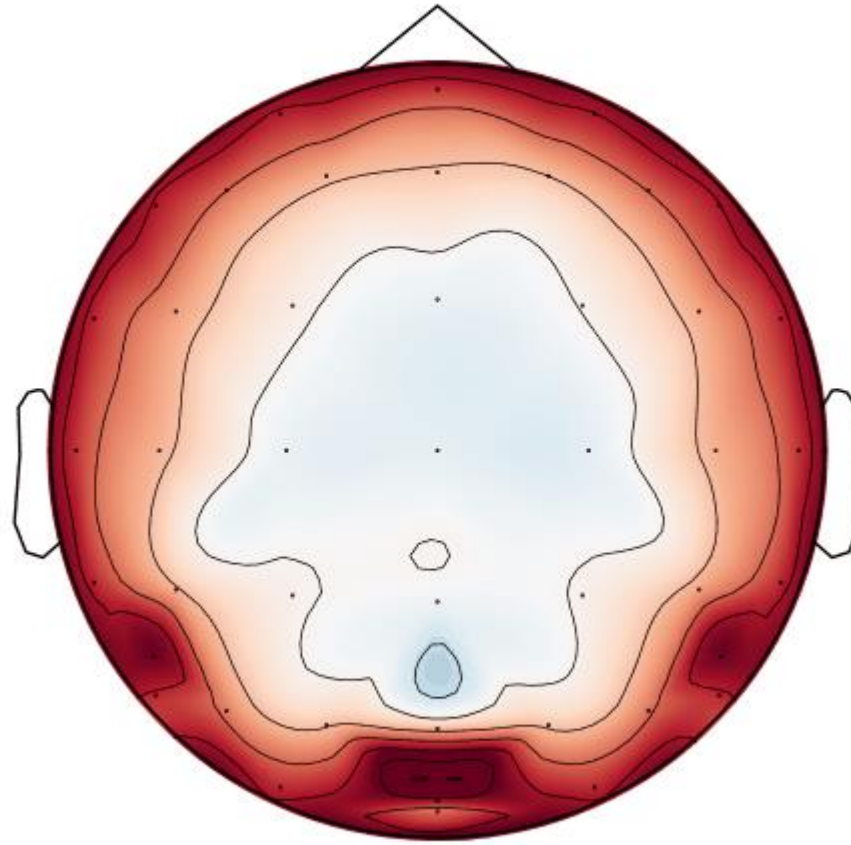


EEG Topomaps for different set of electrodes



EEG Topomaps for 64 electrodes

Topomap at $t = 0$ ms



Conclusion

- Integrated Wilson–Cowan dynamics with connectome-defined node coordinates.
- Computed a forward lead field using a spherical head model and Biosemi64/occipital montage.
- Simulated realistic EEG waveforms from projected excitatory activity.
- Developed animated scalp maps to visualize spatiotemporal EEG patterns.
- Validation showed physiologically plausible features (eg: 1/f spectrum).
- Future work includes refining head models and expanding sensor configurations.

Thank you!



