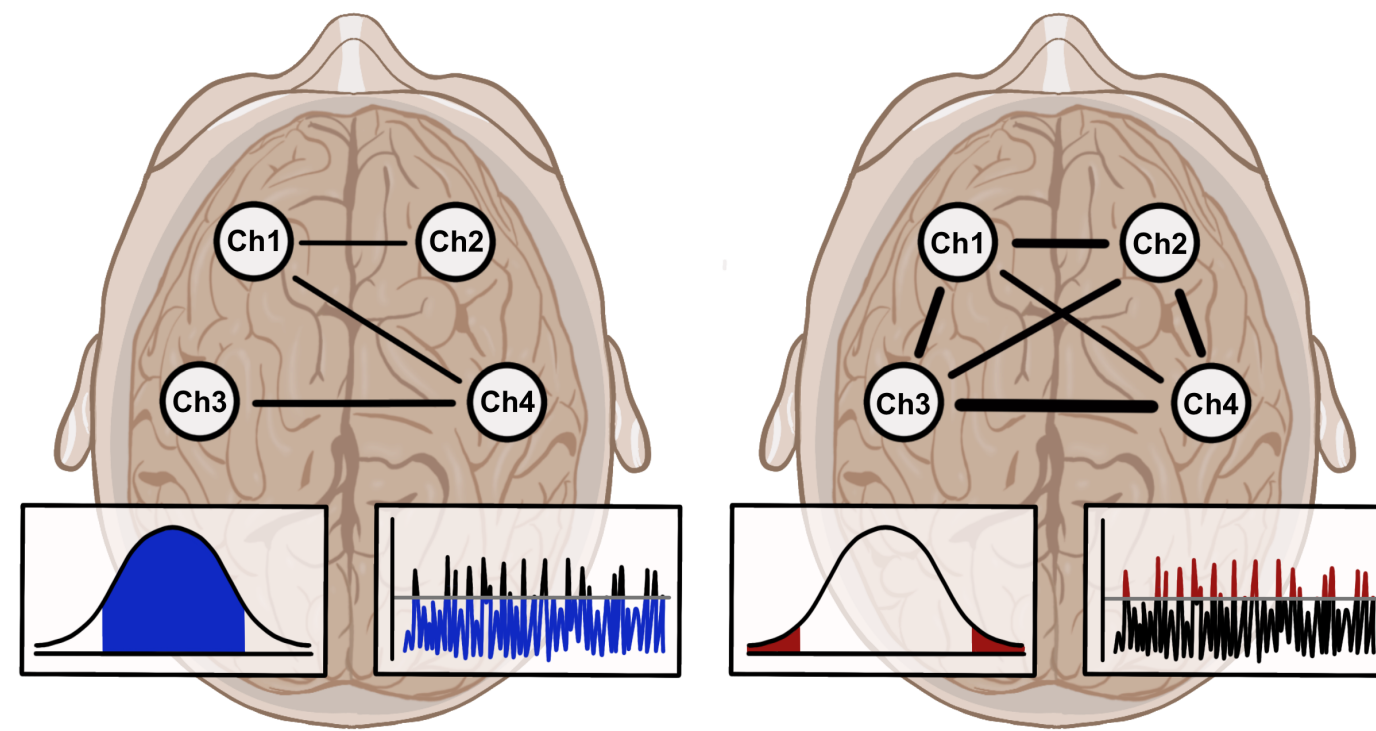


Abstract

► We develop a new conditional **extremal** dependence measure that captures the strength of relationship (beyond linear dependence) between two random variables in a **tail region** given other variables in a network.

Advantages:

- (i) model-free;
- (ii) computed in time $O(n \log n)$;
- (iii) inference using a nonparametric test.



► **Motivation:** Difference between **classical** and **extremal** brain connectivity.

► **Caveat:** We define extremal dependence as the dependence between variables in a tail region which is not equivalent to asymptotic dependence, i.e., $\lim_{u \rightarrow 1} \mathbb{P}(Y > F_Y^{-1}(u) | X > F_X^{-1}(u)) > 0$.

Conditional Extremal Dependence

Let $(X, Y, Z)^T$ be a random vector and consider a quantile level $u \in [0, 1]$. Define the new triple $(X^*, Y^*, Z^*)^T$ as $(X, Y, Z)^T | X > F_X^{-1}(u)$, where $F_X^{-1}(u)$ is the u -quantile of X . For a specified quantile level u , the **conditional extremal dependence** between X and Y given Z is measured by

$$\xi_u(X, Y | Z) = \frac{\int \mathbb{E} [\text{Var}(\mathbb{P}(Y^* \geq t | X^*, Z^*) | Z^*)] d\mu^*(t)}{\int \mathbb{E} [\text{Var}(1_{\{Y^* \geq t\}} | Z^*)] d\mu^*(t)} \in [0, 1],$$

where $\mu^*(\cdot)$ is the distribution of $Y^* = Y | X > F_X^{-1}(u)$ [extends [1]].

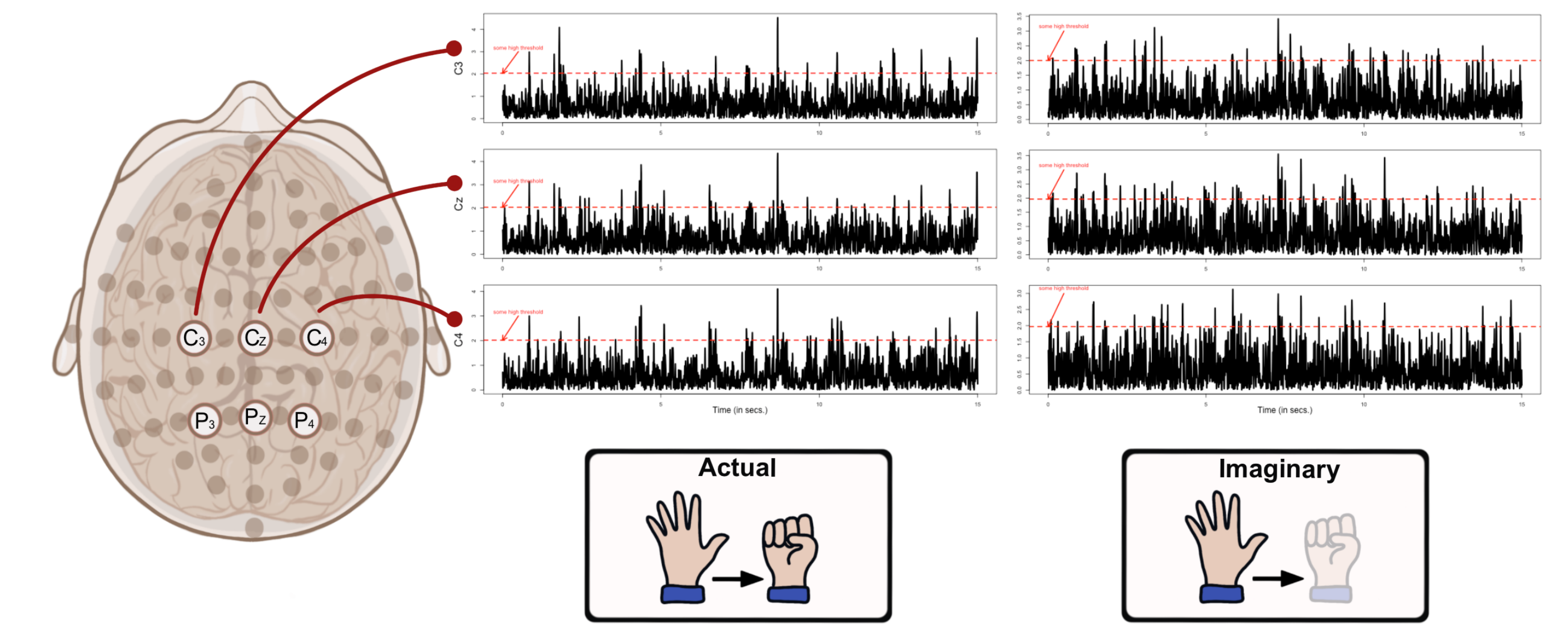
- $\xi_u(X, Y | Z) = 0$ if and only if X^* and Y^* are independent given Z^* , and $\xi_u(X, Y | Z) = 1$ if and only if Y^* is a function of X^* given Z^* .
- $\xi_u(X, Y | Z)$ is **not** symmetric.
- **Remark:** We say $(X^*, Y^*, Z^*)^T$ is in the **tail region** C_u , which is defined by X exceeding a high threshold.

Motor Movement/Imagery EEG Analysis

Data: EEG recordings (at 160 Hz) during performance of motor-related tasks from ten volunteers.

Tasks: (i) actual movement (open/close fist), and (ii) imaginary action.

Selected Channels: C3, Cz and C4 – body sensory and motor functions; P3, Pz and P4 – facilitate accurate upper limb movements.



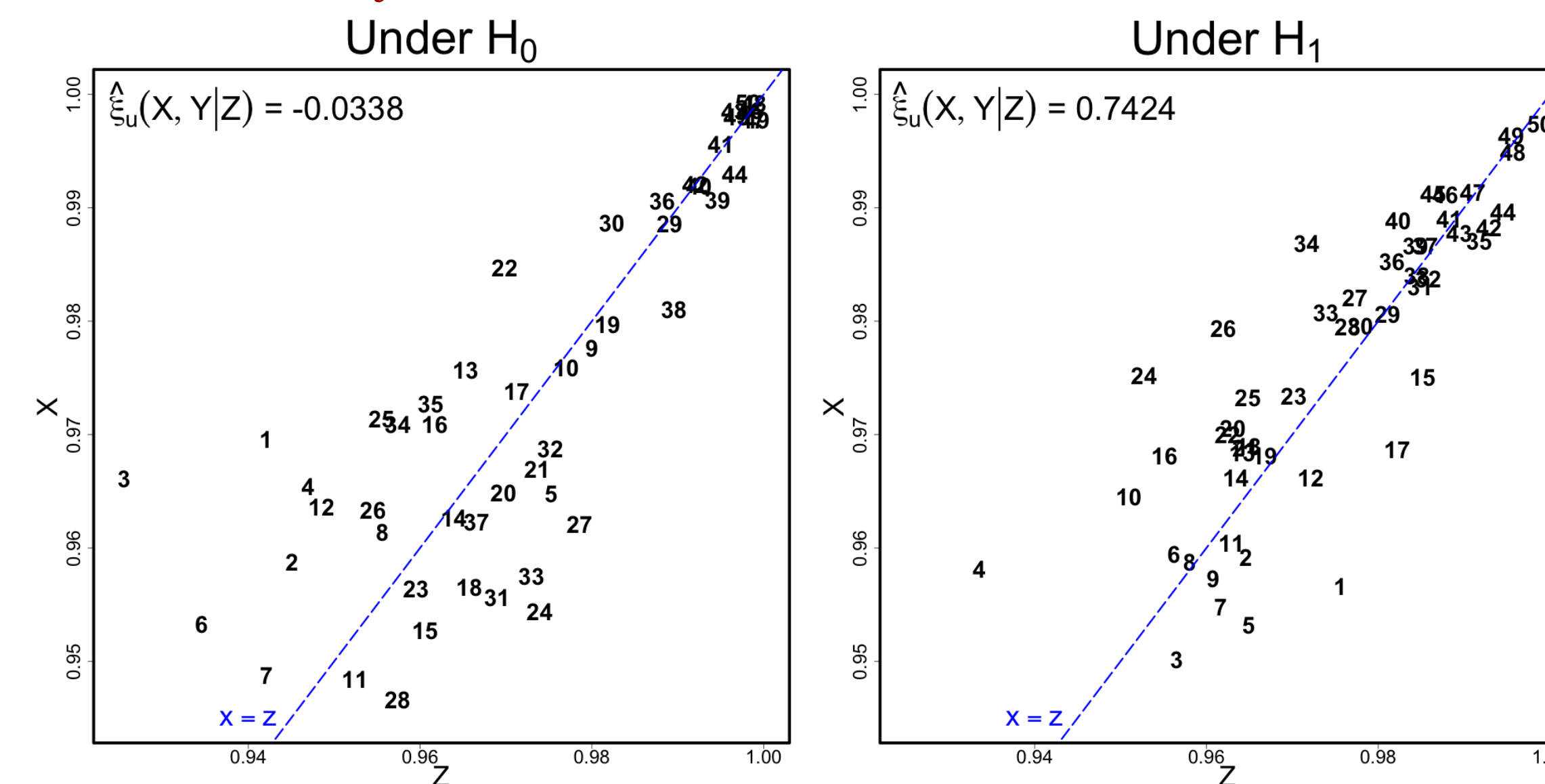
Nonparametric Inference on Conditional Extremal Dependence

Estimation: Suppose data consist of iid triples $(X_i, Y_i, Z_i)^T, i = 1, \dots, m$. For a quantile level u , define new obs. $(X_j^*, Y_j^*, Z_j^*)^T$ such that $X_j^* > F_X^{-1}(u)$ for all $j = 1, \dots, n$. The estimator of $\xi_u(X, Y | Z)$ is

$$\hat{\xi}_u(X, Y | Z) = \frac{\sum_{j=1}^n (\min\{R_j, R_{M(j)}\} - \min\{R_j, R_{N(j)}\})}{\sum_{j=1}^n (R_j - \min\{R_j, R_{N(j)}\})},$$

where R_j is the rank of Y_j^* , $N(j)$ is the index k such that Z_k^* is the nearest neighbor of Z_j^* on \mathbb{R} , and $M(j)$ is the index k such that $(X_k^*, Z_k^*)^T$ is the nearest neighbor of $(X_j^*, Z_j^*)^T$ in \mathbb{R}^2 .

Behavior of Ranks R_j :



Test for Independence in the Tail Region C_u :

Suppose we wish to test the null hypothesis that X and Y are independent in the tail region C_u given Z , i.e.,

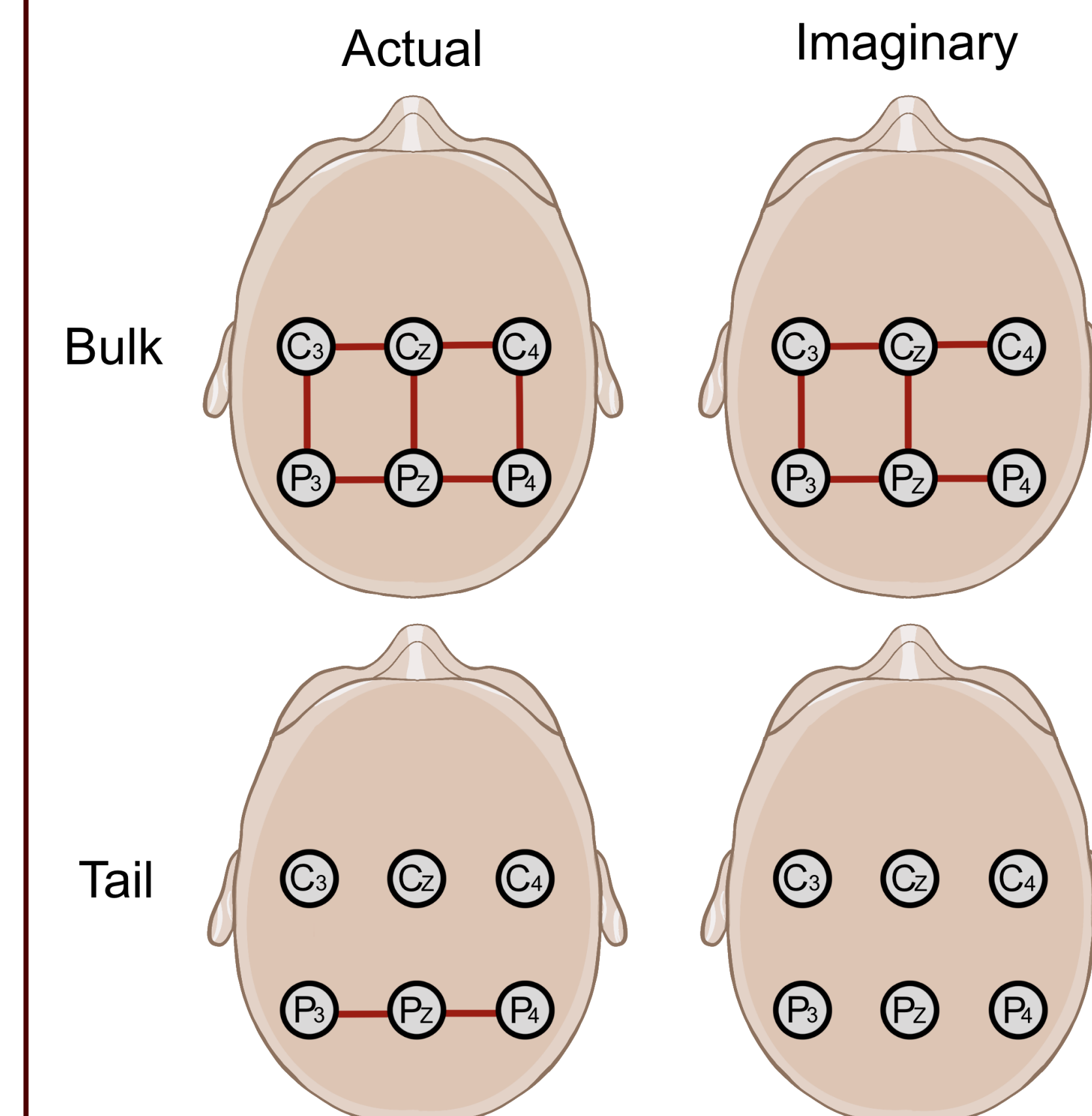
$$H_0 : \xi_u(X, Y | Z) = 0 \text{ vs. } H_1 : \xi_u(X, Y | Z) \neq 0.$$

We construct the empirical null distribution of the estimator $\hat{\xi}_u(X, Y | Z)$ by simple “reshuffling” of adjacent ranks.

Outline:

1. Generate independent random permutations of $R_{M(j)}$ and $R_{N(j)}$.
2. Calculate $\hat{\xi}_u(X, Y | Z)$ based on the reshuffled ranks and denote by $\hat{\xi}_u^{(b)}(X, Y | Z)$.
3. Compute **p-value** as $|\{\hat{\xi}_u^{(b)}(X, Y | Z) \geq \hat{\xi}_u(X, Y | Z)\}|/B$, where B is the number of reshuffles considered.
4. Given significance $\alpha \in (0, 1)$, reject the null hypothesis if the **p-value** is less than α . Otherwise, do not reject H_0 .

Estimated Brain Connectivity Network:



Novel Findings:

- There is a **direct link** (represented by a **red line**) between two channels when their signals are conditionally dependent given all other channels in the network
- **Bulk:** Similar **systematic** brain connectivity for the two tasks.
- **Tail:** **Prominent direct links** in the parietal region when performing actual movement while **conditional independence** during imaginary action.

Future Work:

- Develop a **causal** extremal measure, e.g., $\xi_u(X_{t-k}, Y_t | Y_{t-l})$.

References:

- [1] Azadkia, M. and Chatterjee, S. *simple measure of conditional dependence*. The Annals of Statistics 49 (2021): 3070–3102.
- [2] Chung, J. W., et al. *Beta-band activity and connectivity in sensorimotor and parietal cortex are important for accurate motor performance*. Neuroimage 144 (2017): 164–173.

