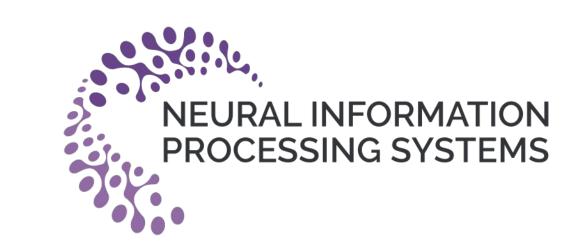
# Exploring latent networks in resting-state fMRI using voxel-to-voxel causal modeling feature selection



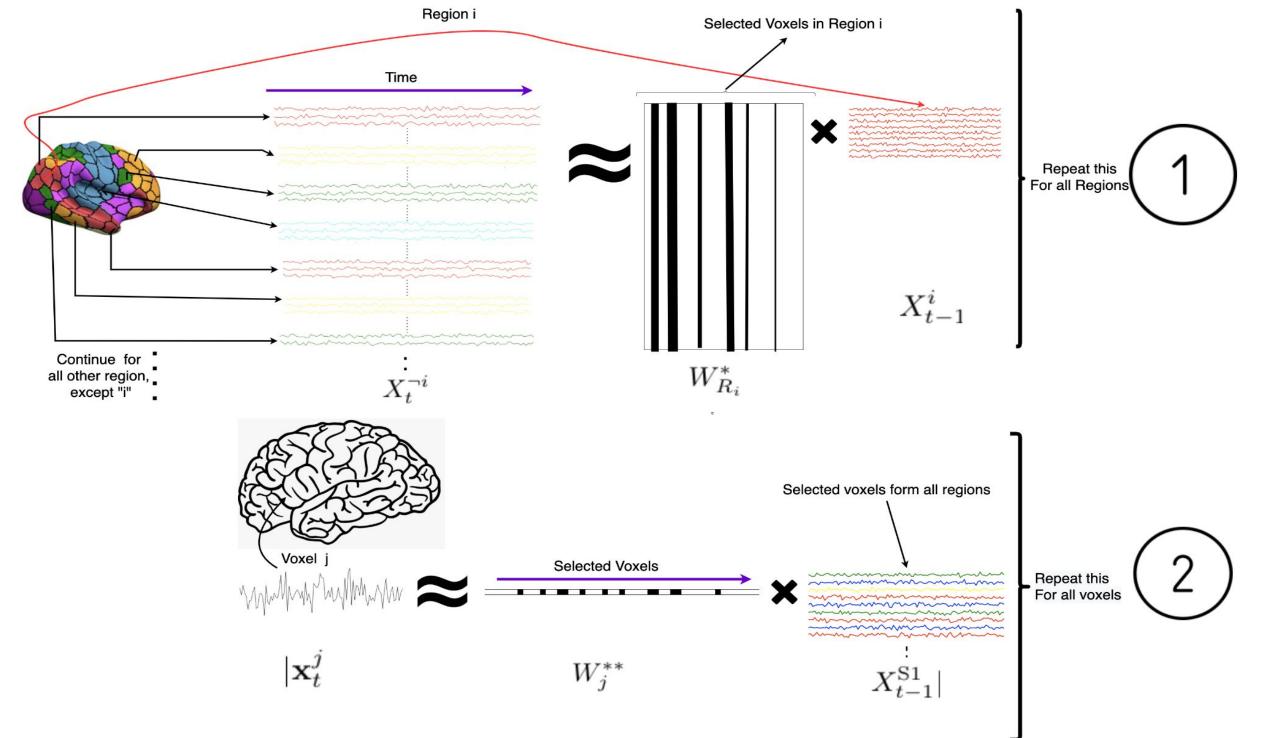


#### Goal

- Model the rs-fMRI for all cortical grey-matter voxels using a subset of predictive voxels.
- Find latent networks within the selected voxels.

## Methodology (Stage 1 and 2)

Fig. I: Divide and conquer sparse linear modeling approach



## Methodology (Stage 3)

- (Stage 3) Apply ICA on stage 2 voxels and project each source back using predictive coefficients.
  - Align to common space (MNI152) and blur to deal with cortical misalignment.

#### Results

 $W \in \mathbb{R}^{(V-V_i) \times V_i}$ 

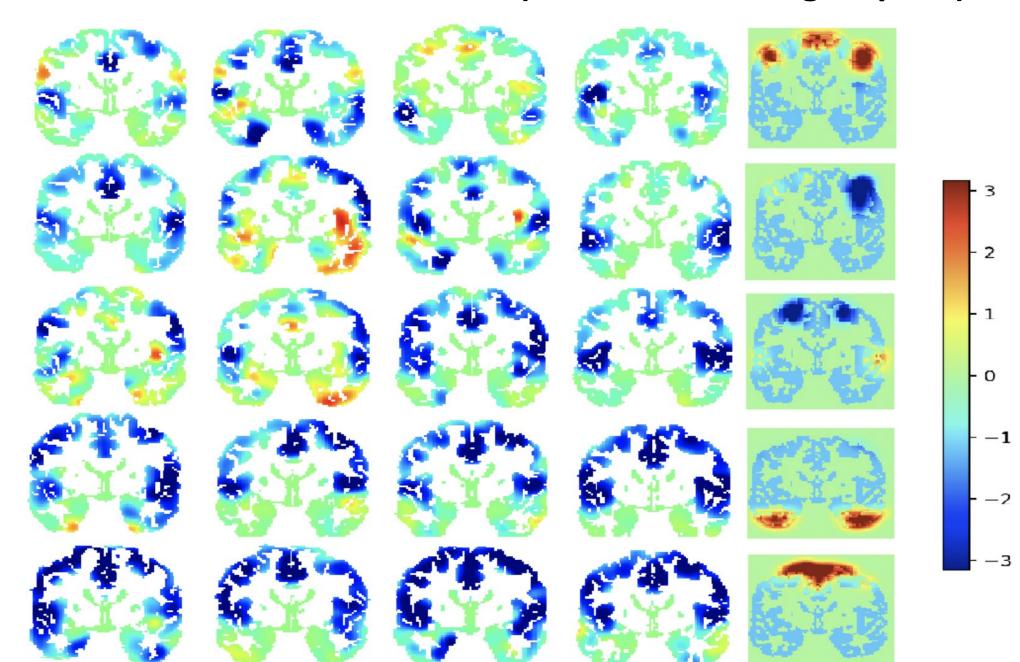
- Method finds unique ICs that high similarity across multiple subjects but obtained from group-based ICA.
- that group-based ICA is not able to find.

 $\arg\min \|X_t^{-i} - WX_{t-1}^i\|_F^2 + \lambda_{2,1} \|W\|_{2,1},$ 

- low similarity with ICs
- (Fig. III) Our analysis is able to find common latent networks across subjects
  - Fig. III: Slices of ICs from cluster D (last column are group ICs)

Fig. II: Hierarchical clustering of subject ICs by inter-subject

similarity and similarity to group ICs (column labeled 24)



### **Future Work**

We are now testing whether these unique IC patterns are meaningful in distinguishing healthy versus non-healthy subjects.

- (Stage 1) Find voxels within each region.
  - $\circ$  For each of 1000 regions, use a  $\ell_{21}$ -norm penalized linear causal model to predict the activity at the next time step of all other regions.
    - Take the union over predictive voxel subsets for each region.
  - (Stage 2) For each voxel, find a non-redundant set of voxels from stage 1 by applying an ℓ₁-norm penalized linear model (LASSO).

$$\mathbf{w}_{j}^{**} = \underset{\mathbf{w} \in \mathbb{R}^{V_{S1}}}{\min} \|\mathbf{x}_{t}^{j} - (X_{t-1}^{S1})^{\mathsf{T}} \mathbf{w}\|_{2}^{2} + \lambda_{j} \|\mathbf{w}\|_{1},$$