

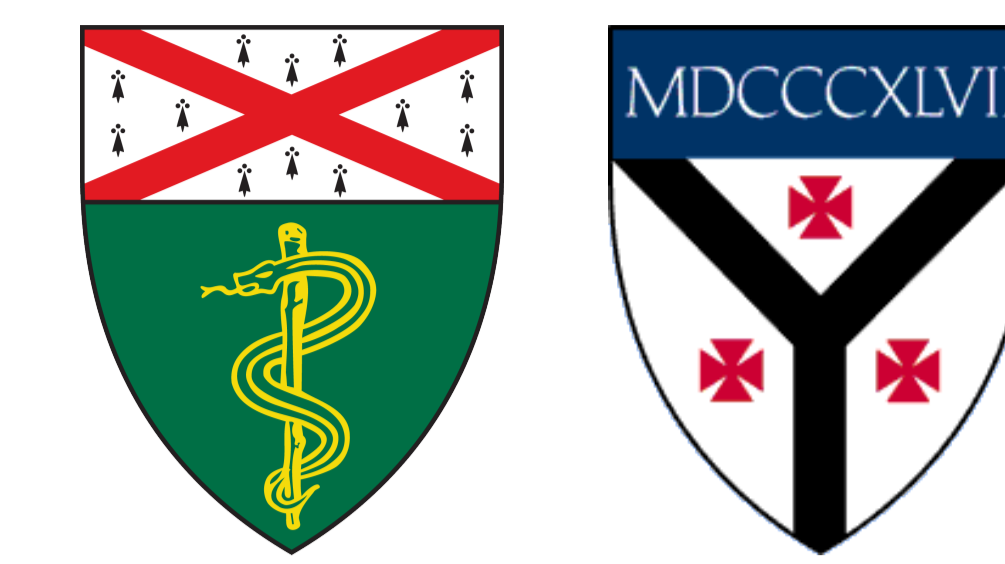
# Predictive models of IQ from functional connectivity data

## may not be sex specific

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### Introduction

- Functional connectomics is derived from **functional magnetic resonance imaging (fMRI)** for individual and group differences in brain organization.
- Predictive models** have long been used to predict behavioral measures.
- But the question is if these models are sex specific? If a model built on one group is generalizes to the other?

### Sex Specific Pipeline of IQ

- fMRI data consisting of  $k$  tasks acquired from two groups of participants (i.e., male vs female);
- Parcellate the brain into  $N$  nodes;
- Average **timeseries** for each node;
- Generate connectomes  $\mathbf{x}_i \in \mathbb{R}^{N \times N}$  for **each task** using the time series (e.g., we have 9 different connectome for a person in HCP);
- Treat these edges as features;
- Train a predictive model on one sex, and test on the same group.
- Train a predictive model in one sex and test on the opposite sex.
- Train a predictive model that classifies participants based on sex.
- visualize the models;

### Sex Specific Models

ID	Collection	#male	#female	size	age	#tasks
HCP	Human Connectome Project	241	274	515	28 ± 3.98	9

Table 1: Characteristics for the HCP and PNC datasets.

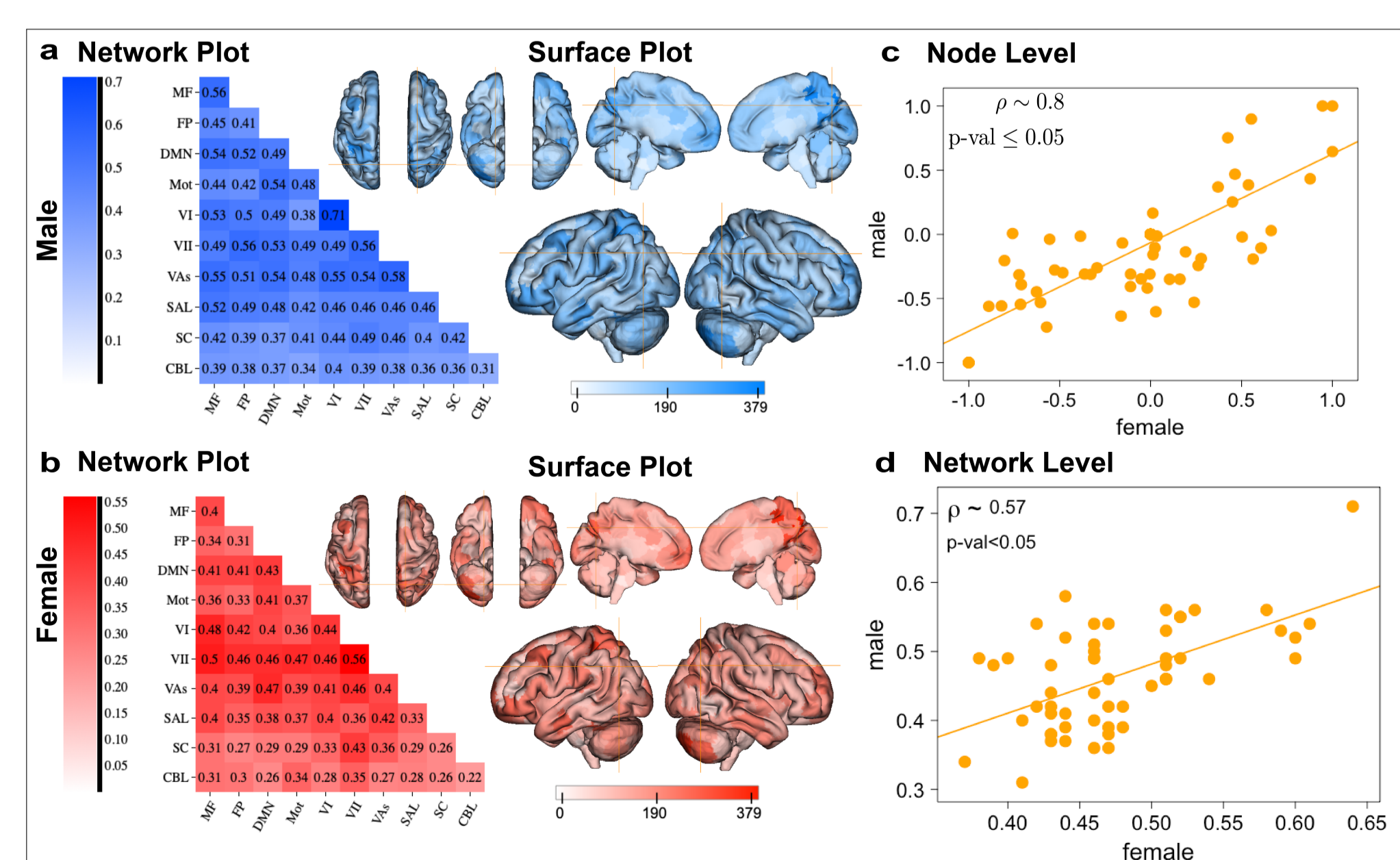


Figure 1: Network and degree plots and correlation of sex specific models.

### Sex Specific Predictive Model Pipeline

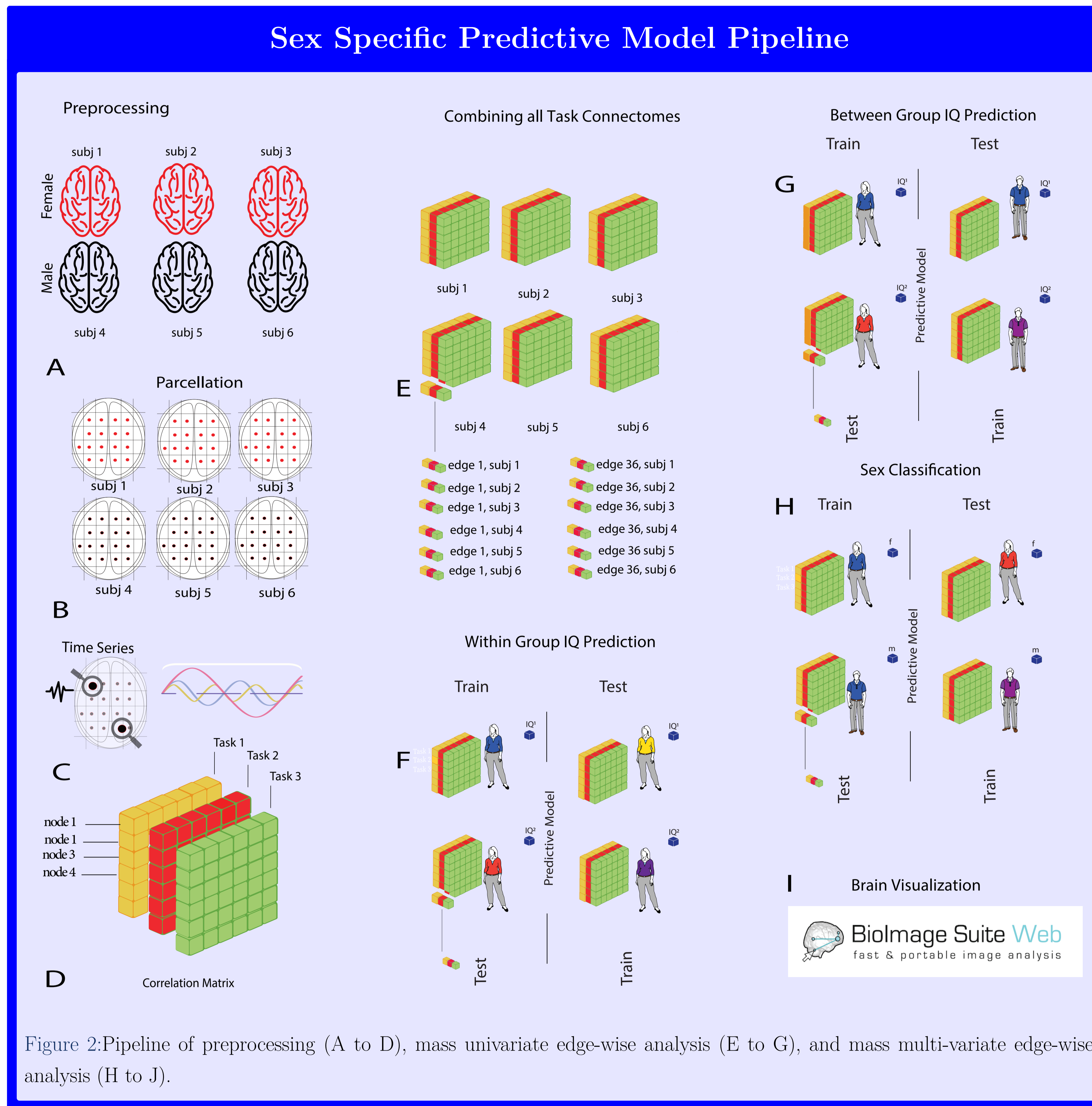


Figure 2: Pipeline of preprocessing (A to D), mass univariate edge-wise analysis (E to G), and mass multi-variate edge-wise analysis (H to J).

### Cross-sex Prediction

- We divide participants into two groups with equal size (e.g., 200 in HCP).
- We train ridge connectome based predictive model (rCPM) on each group
- Then we test the models on opposite side:

$$\min_{\beta} (Y - \beta X)^2 + \lambda \|\beta\| \quad (1)$$

where  $Y \in \mathbb{R}^N$  is the vector of IQ measures,  $\beta \in \mathbb{R}^K$  is the coefficient vector and  $X \in \mathbb{R}^{K \times N}$  is the feature matrix.

### Cross-sex Prediction

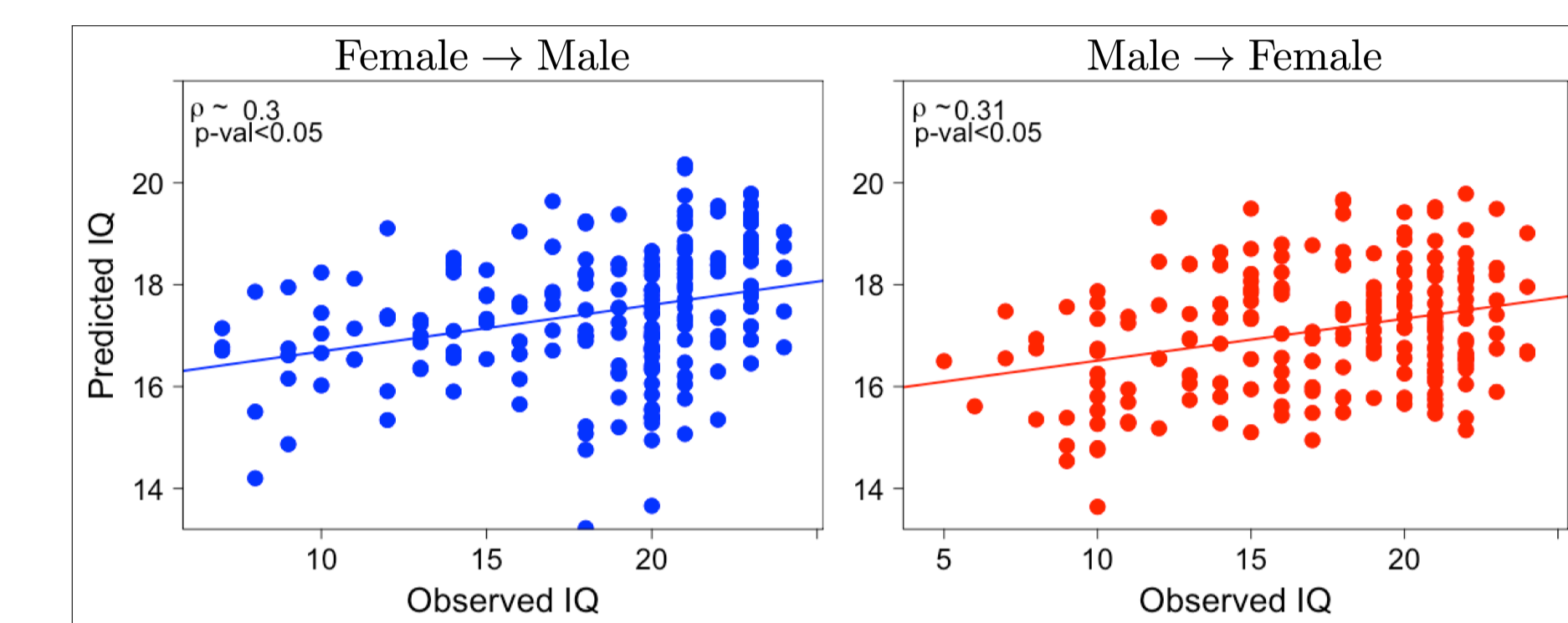


Figure 3: Pearson correlation of predicted and observed IQ.

### Influence of Brain States

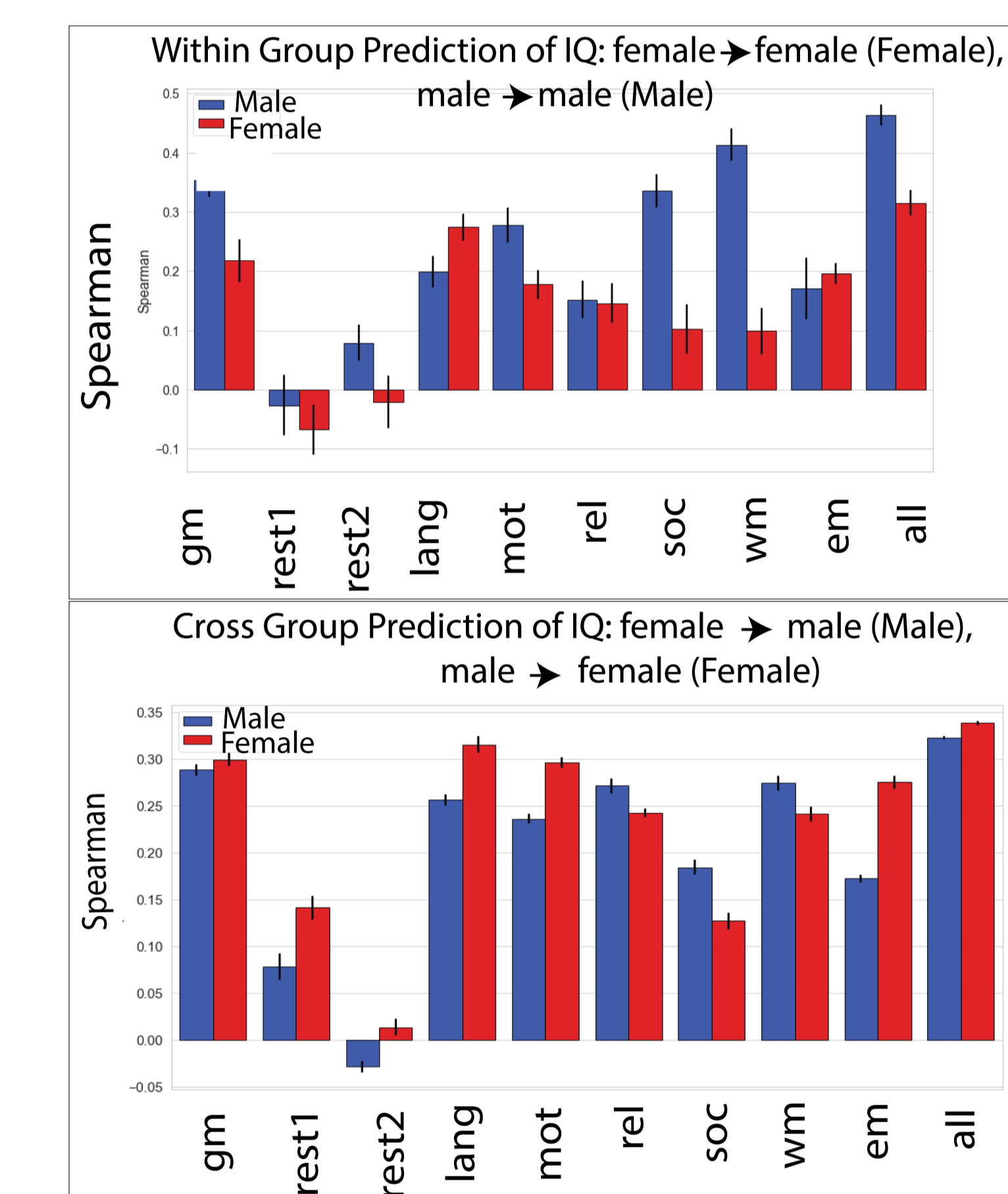


Figure 4: Influence of difference brain states within group (top) and between group (bottom).

### Classification of Sex

	Predicted Female	Predicted Male
Observed Female	266	8
Observed Male	19	222
	285	230

### Conclusion

We investigated sex specificity of predictive models. Experimental results show that predictive models may not be sex specific in combined connectomes and specific tasks.

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