## **Integrating Multimodal Neuroimaging Features to Predict** Working Memory and Psychiatric Disability

#1124

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## Background

- Working memory capacity (WMC) is positively correlated with higher order cognitive ability<sup>1,2</sup> and negatively correlated with psychiatric disability<sup>3,4</sup>
- A variety of structural and functional neuroimaging measures have been shown to correlate with individual differences in WMC
- •Most studies that investigate WMC only look at one type of neuroimaging measure, or within one psychiatric population
- Using a large sample (N=169) and an ensemble machine learning framework aimed at harnessing the informativeness of different neuroimaging feature classes<sup>5,6</sup>, we predicted individual differences in working memory task performance, trait WMC, and psychiatric disability

# Analytic Approach: Feature Space

• Functional data parcellated into 400 region Schaefer atlas and restricted to regions that show high > low load effects in any task measure (FDR corrected)



- 84 regions across Control (37), Visual (15), Dorsal Attention (15), Default (9), Salience/Ventral Attention (8) and Somatomotor (1) networks identified
- Despite a high degree of spatial overlap in load effects across task fMRI

## **Experimental Design**



### **Behavioral Measures**

#### **Delayed Face Recognition (DFR) Task Performance**

- Delayed-match-to-sample working memory task requiring maintenance of either 1 face (low load) or 3 faces (high load).
- 32 high load and 32 low load trials total (across 4 scanner runs); accuracy averaged into a single DFR Task Performance score

measures, no single region showed load effects in all three measures

## Results



#### Visual WMC

E2D Conten Persistence Representation:

#### **DFR Task Performance**



Content	E2D	
Representation:	Persistence	Demographics



#### Working Memory Capacity (WMC)

 Exploratory Factor Analysis using oblimin rotation on 10 independent working memory tests identified latent factors capturing Visual and Verbal WMC

#### World Health Organization Diability Assessment Scale 2.0

• Scores summed across WHODAS used as index of psychiatric disability

#### **Analytic Approach: Stacked Models** Layer 1 ElasticNet Models Layer 2 LASSO Model Univariate GLM: Encoding Univariate GLM: Delay Univariate GLM: Probe ons uo Content Representation: Delay



 Pearson correlation between actual behavioral outcomes and predicted values from held-out test sets revealed that while stacked models could predict all behavioral outcomes, different neuroimaging features were retained for each model

o DFR Task Performance mostly predicted by measures from task fMRI; Visual WMC was also predicted by structural MRI and resting state FC

ofMRI features from DFR task that reflect maintenance of content over the delay period were retained in the model predicting WHODAS

# **Conclusions and Future Directions**

Machina learning models with multimodal feature stacking were able to



significantly predict across-subject variance in all three beha	vioral outcomes
<ul> <li>Diverse measures (including fMRI pattern similarity and functional connectivity), not just delay period univariate ( are important for understanding individual differences in</li> </ul>	d task-based GLM contrasts, working memory
<ul> <li>Measures from the scanned DFR fMRI task are retained in psychiatric disability, suggesting potential utility as transdiage</li> <li>Future work will examine prediction of specific psychiatric statements</li> </ul>	models predicting nostic biomarkers symptom classes
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