

# Assessing individual differences with fMRI

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NIH Summer Neuroimaging Course June 29, 2018

fMRI is barely used clinically...
can we change this?

#### **Outline**

- 1. What do we mean by individual differences?
- 2. Why should we care about individual differences?
- 3. How can we study individual differences? (some practical guidelines)
- 4. Where are we going next?

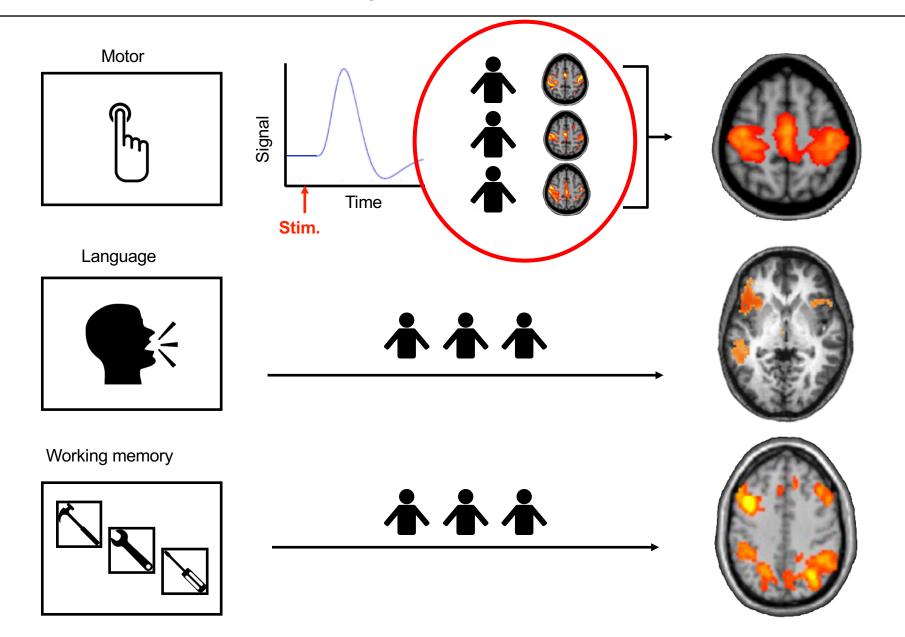


#### **Outline**

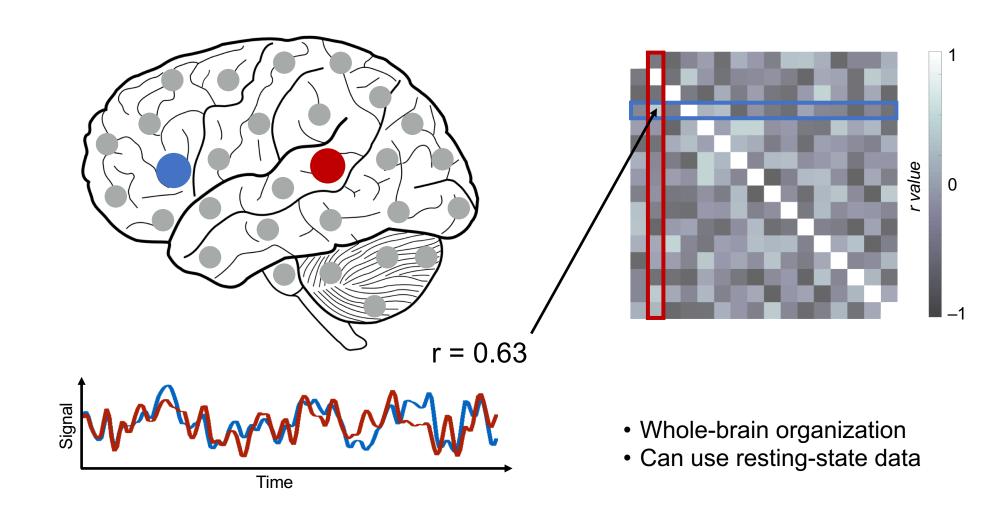
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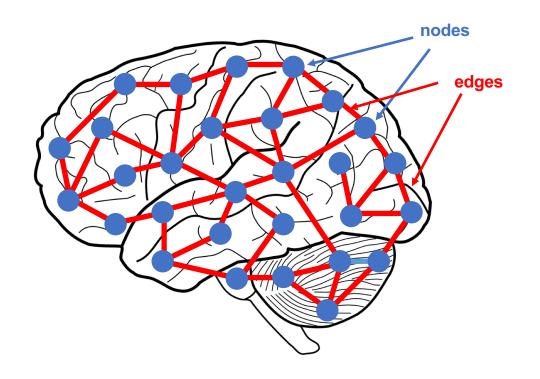
### Traditional fMRI analyses

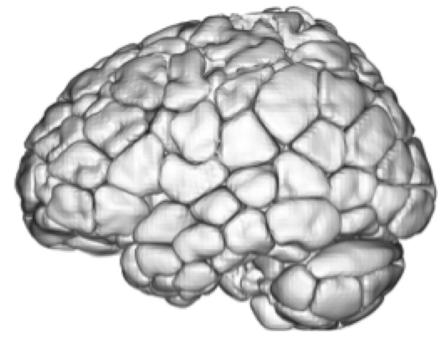


### Functional connectivity



### Brain networks



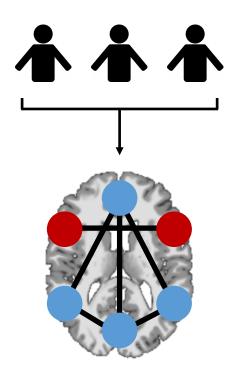


Shen et al., Neurolmage (2013)

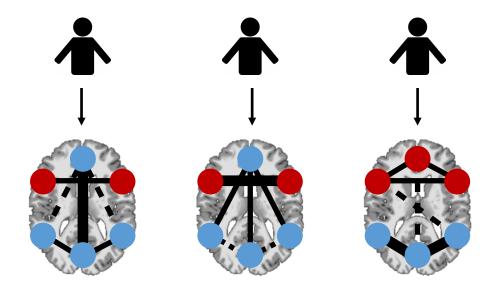
268 nodes = 35,778 edges

#### Individual differences

Group analyses



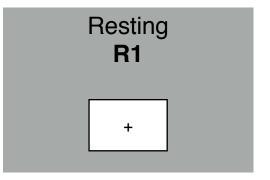
Individual differences

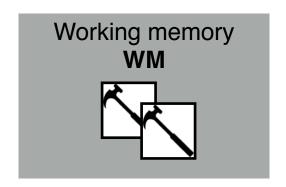


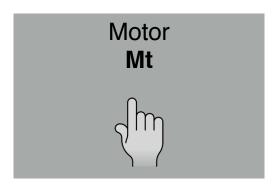
#### **Human Connectome Project**

- 126 healthy subjects (50 sets of twins)
- Age 22-35 years old

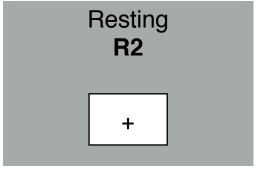
Day 1

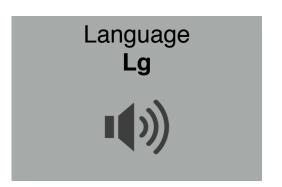


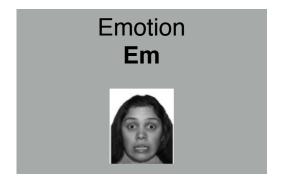




Day 2

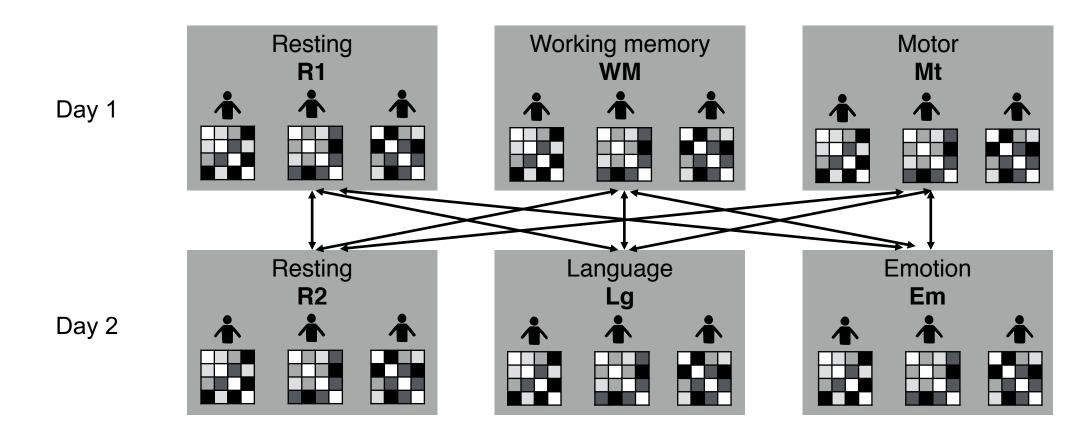


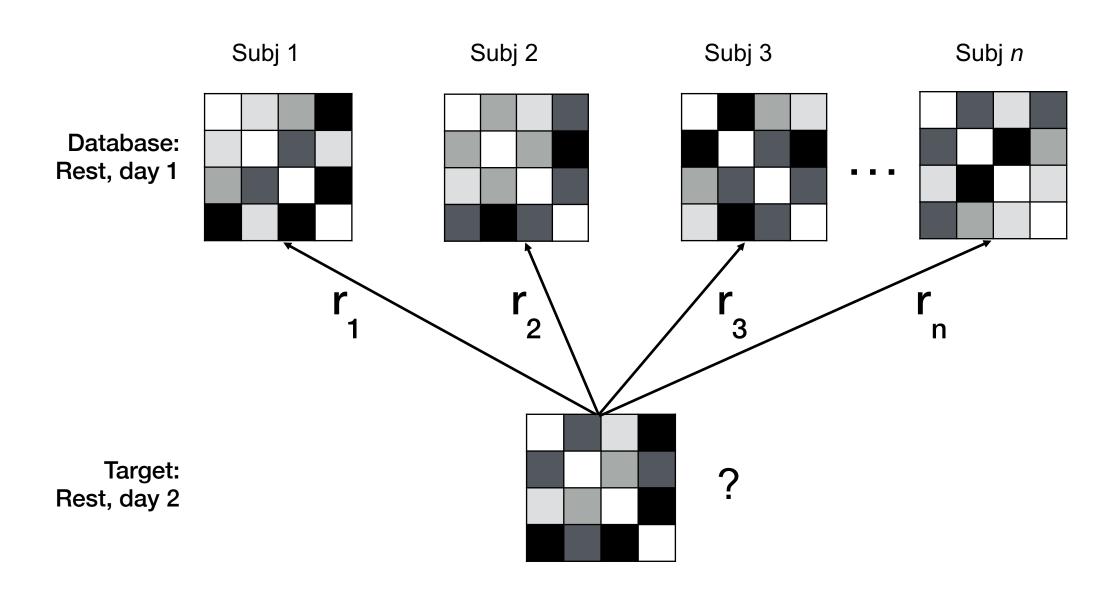


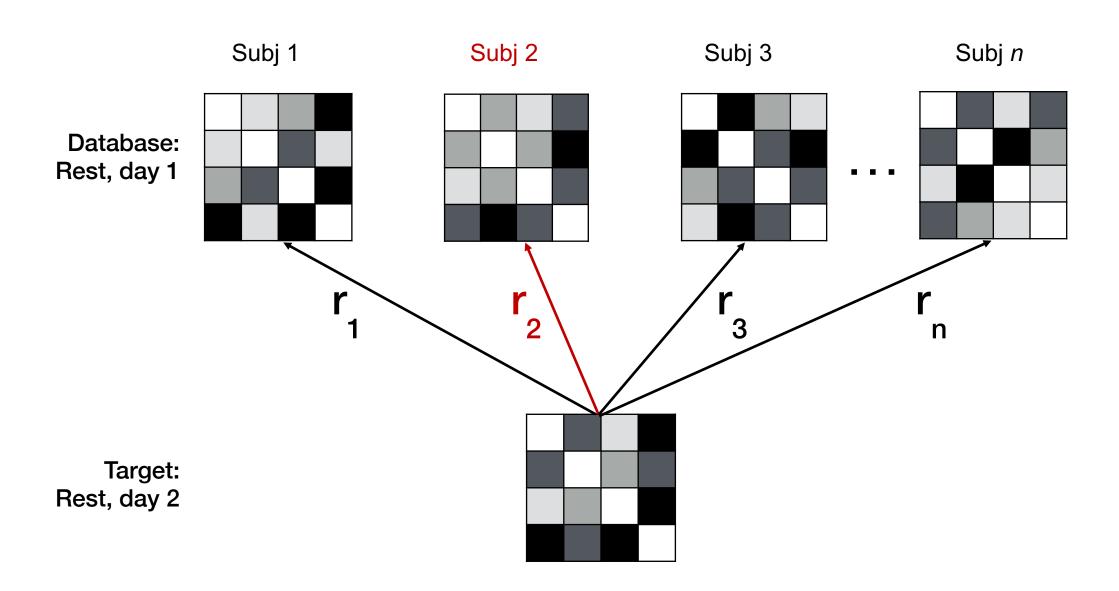


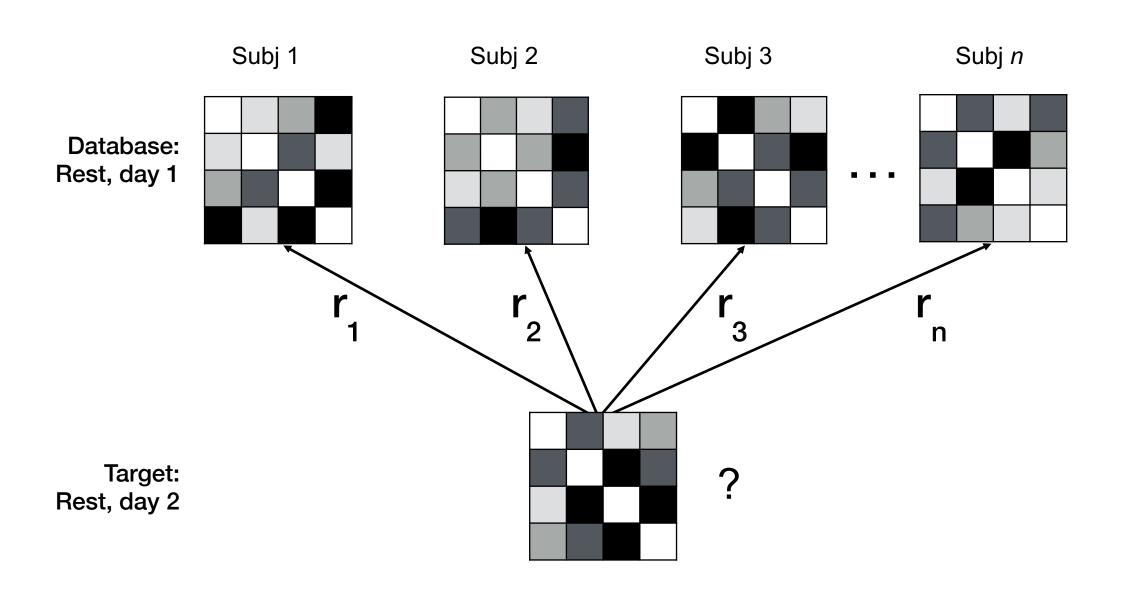
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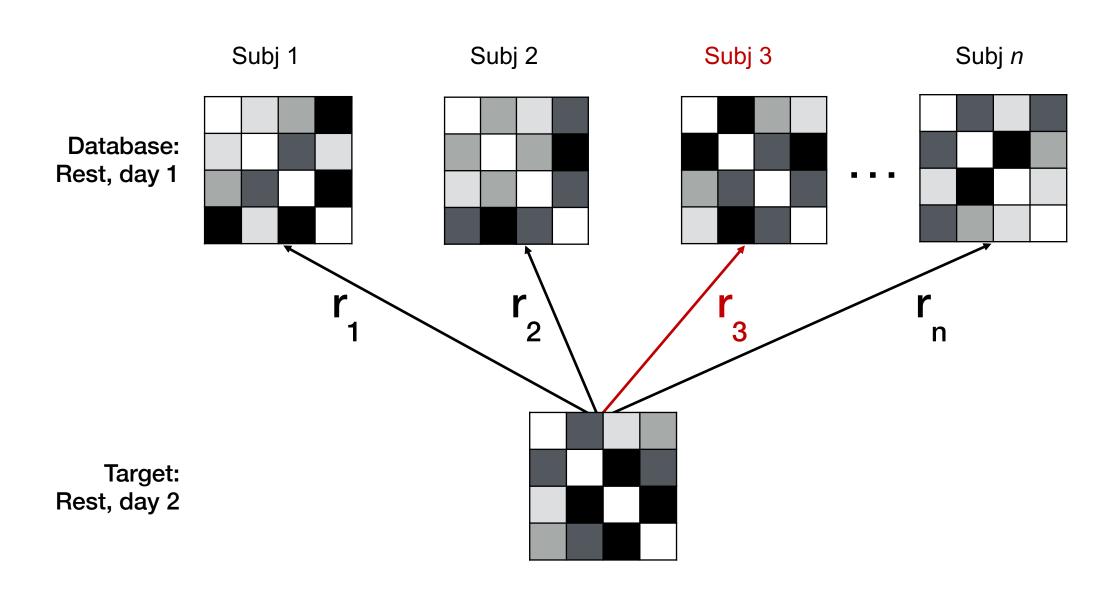
- 126 healthy subjects (50 sets of twins)
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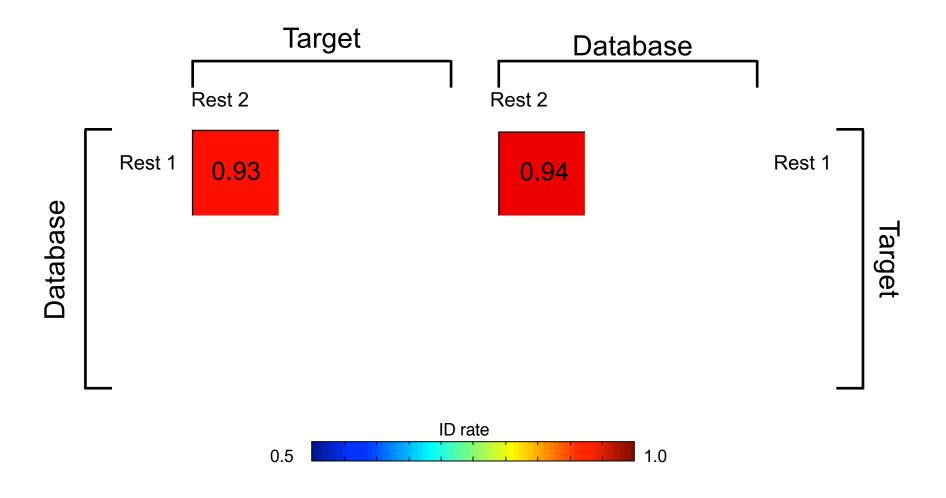








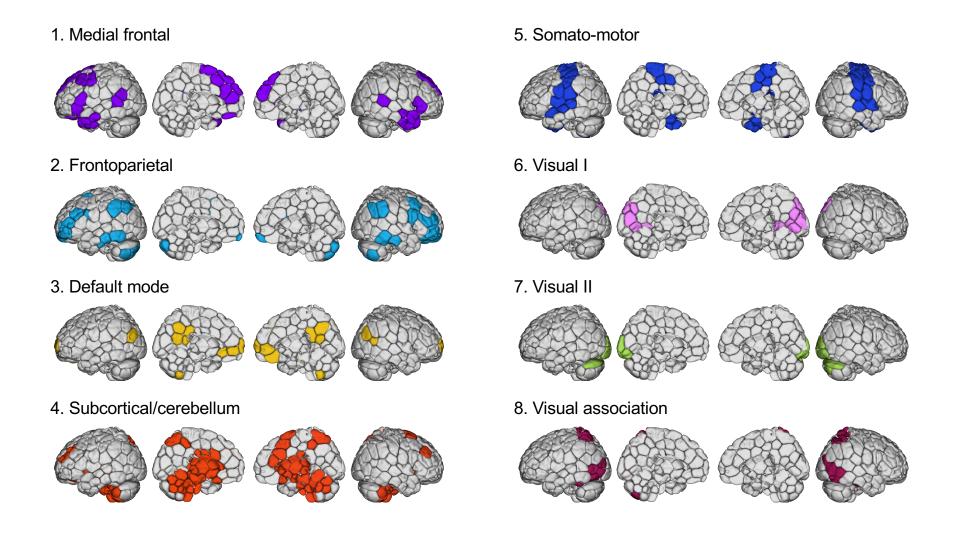
### Identification results



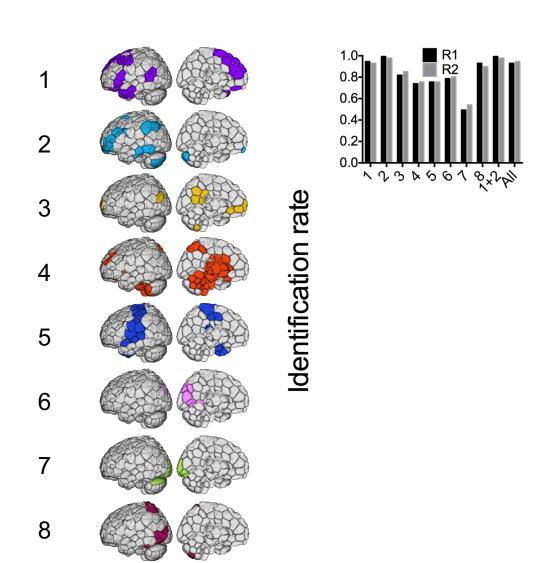
Chance: ~0.008

Finn, Shen et al., Nat Neurosci (2015)

#### Network-based identification



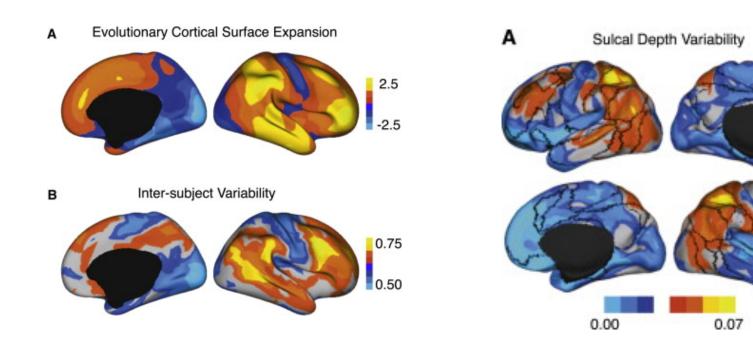
### Network-based identification

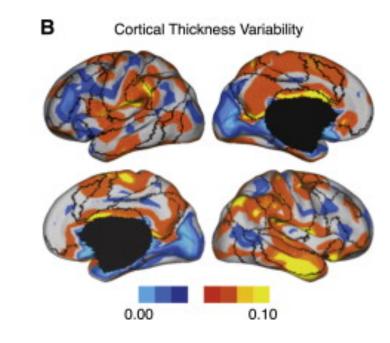


### Localizing individual differences

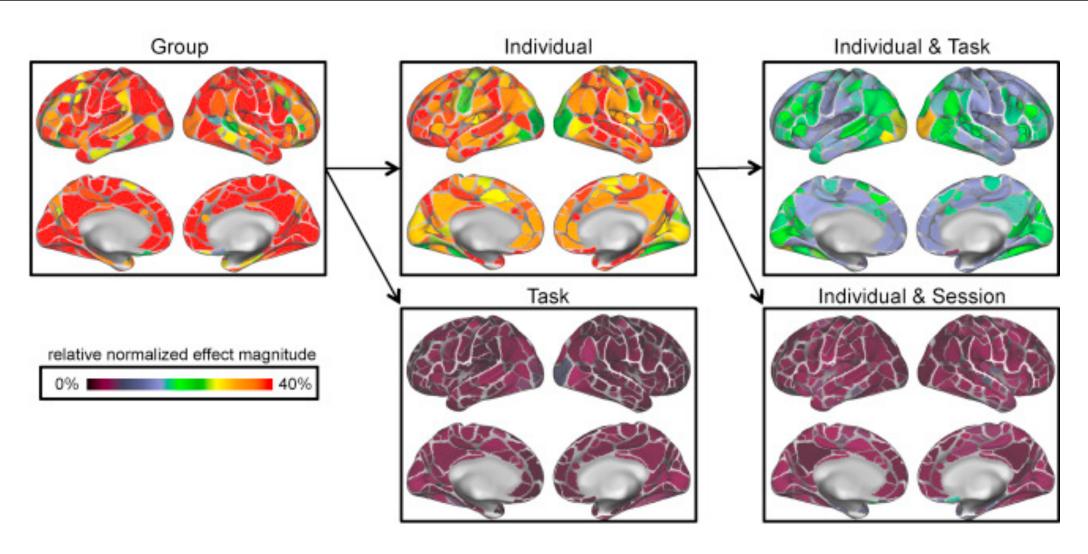
Biggest differences found in most evolutionarily recent regions:

Anatomical differences also play a (large) role:





### Individuals account for the most variance!



Gratton et al., Neuron (2018)

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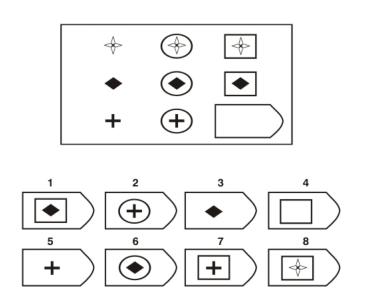
### Why is this important?

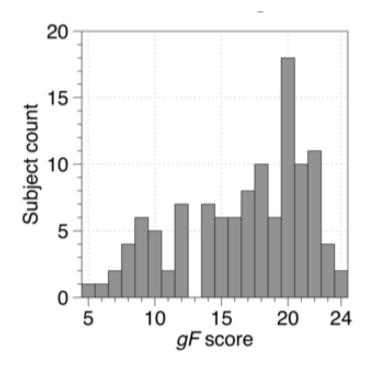
- It's trivial to ID someone based on a structural scan
- Could just be anatomy, other confounds
- How can we prove these differences are meaningful?

# BEHAVIOR!

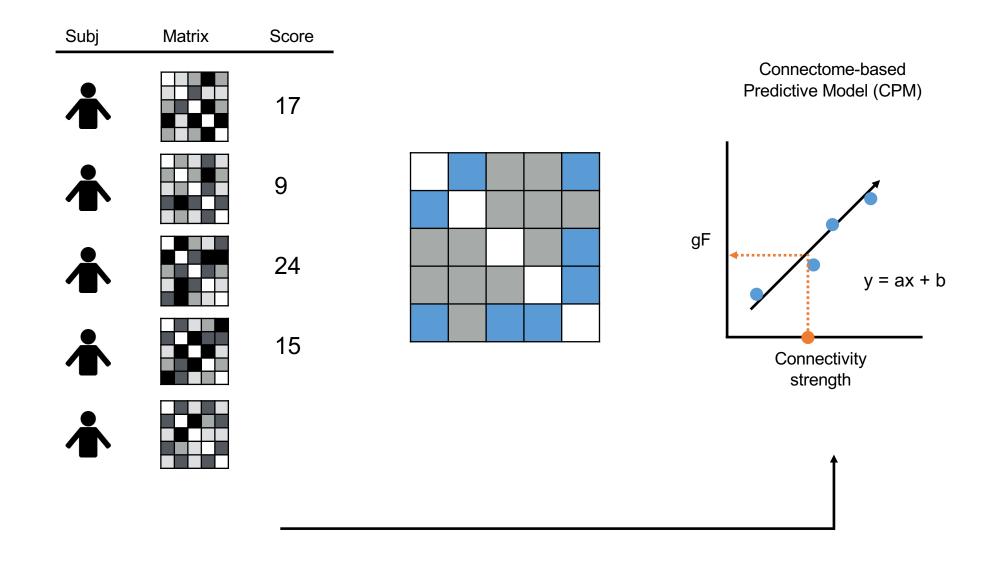
### Predicting fluid intelligence

- ability to discern patterns
- independent of learned knowledge

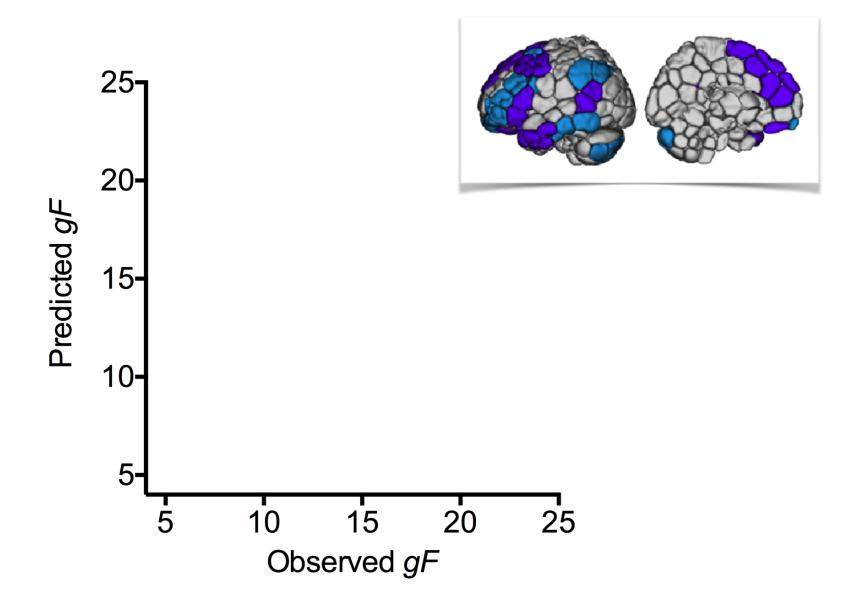




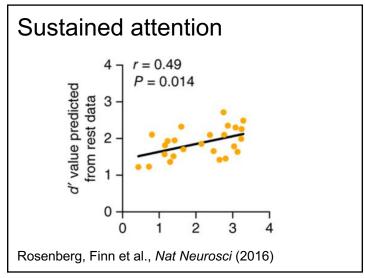
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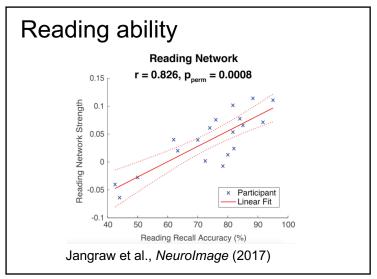


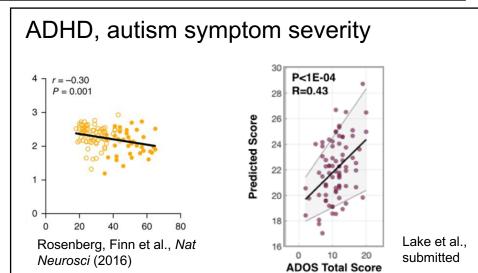
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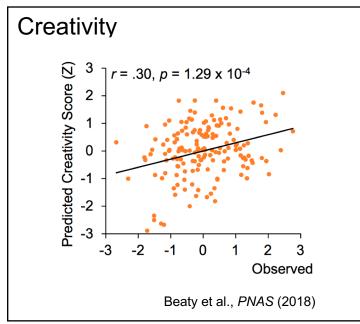


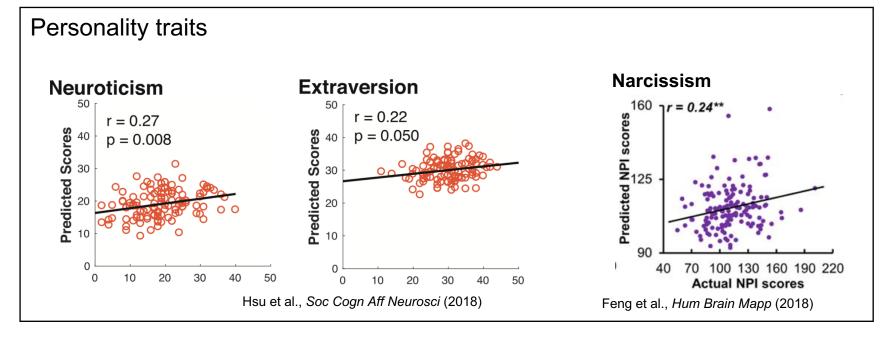
### Predicting other behaviors





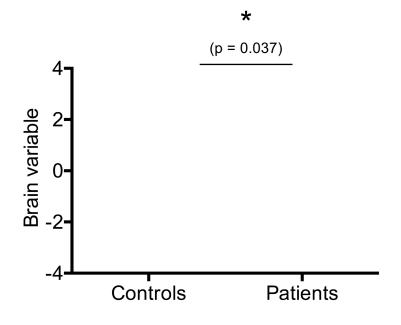




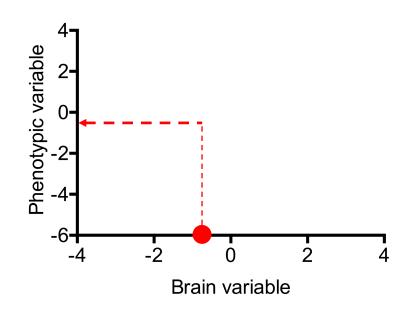


### Translational applications





#### Dimensional approach



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#### **Outline**

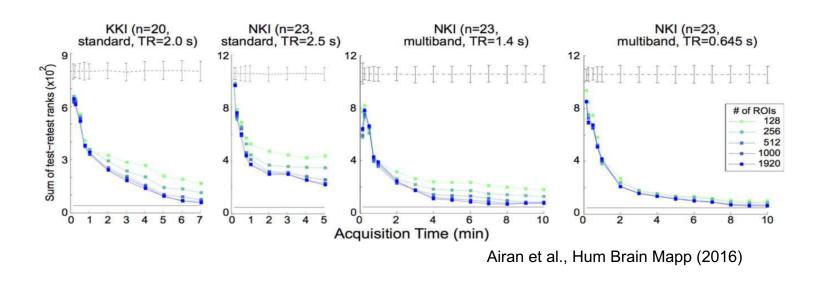
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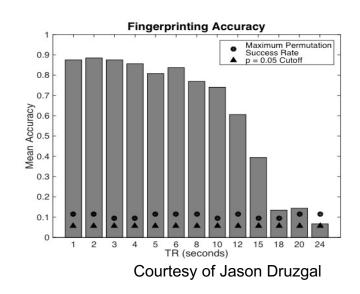


### Q. Do you need HCP-quality data?

#### A. Not really

ID is fairly robust even at more standard spatial & temporal resolutions:





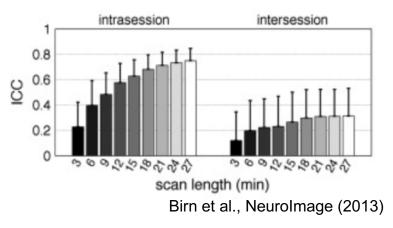
- More nodes → higher identification rate
  - Parcellation method (random vs. functional) did not matter
  - Caution: Higher resolution may amplify effects of anatomical diffs/registration error
  - Parcellations in the 200-300 node range seem like a good compromise

#### Q. What about amount of data?

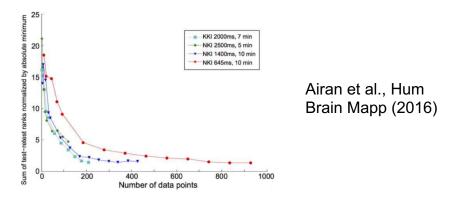
#### A. Scan duration matters!

Longer acquisitions are better:

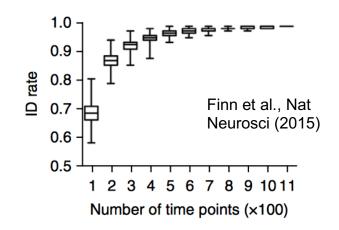
higher reliability within subjects

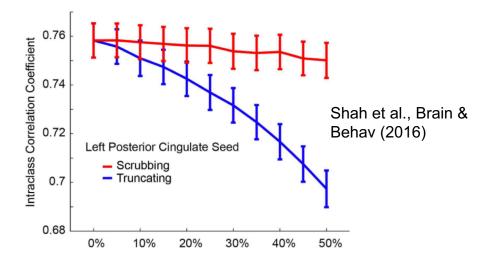


 higher sampling rate (shorter TR) cannot make up for shorter scan duration

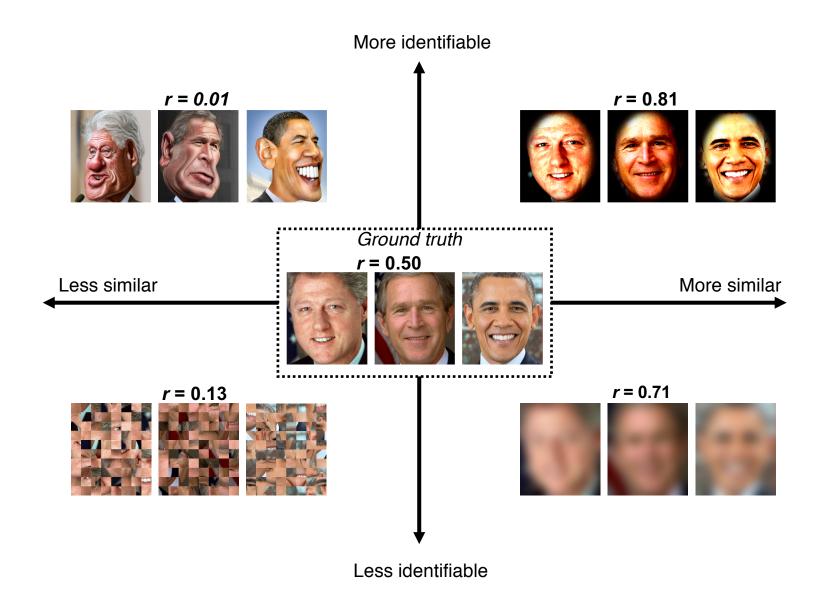


higher identifiability across subjects





#### Q. Does scan condition matter? A. Yes!

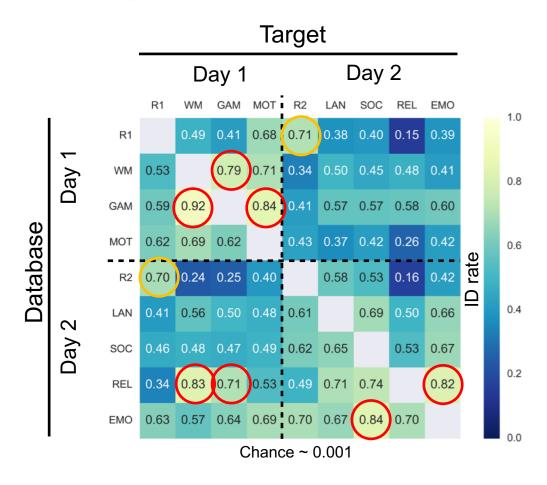


### Q. Does scan condition matter?

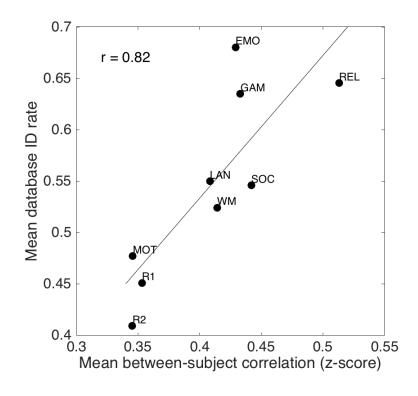
#### A. Yes!

Rest has become the default condition for FC & individual differences, but tasks may increase signal-to-noise

Replicating identification experiments:



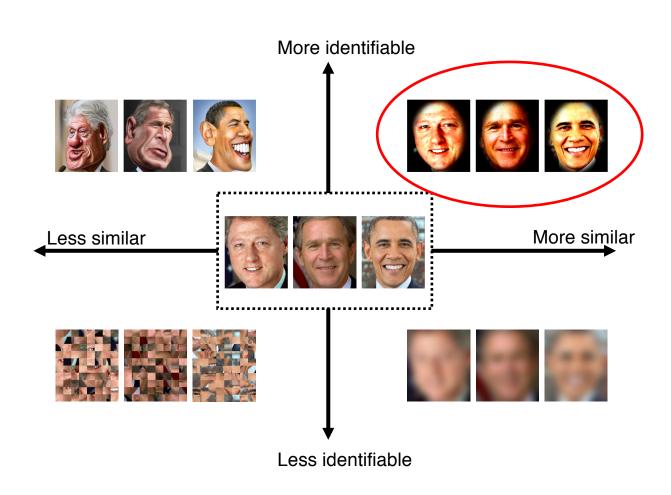
Conditions that make subjects look more similar to one another actually make better databases for identification:



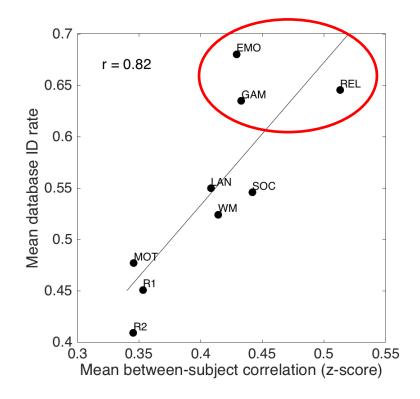
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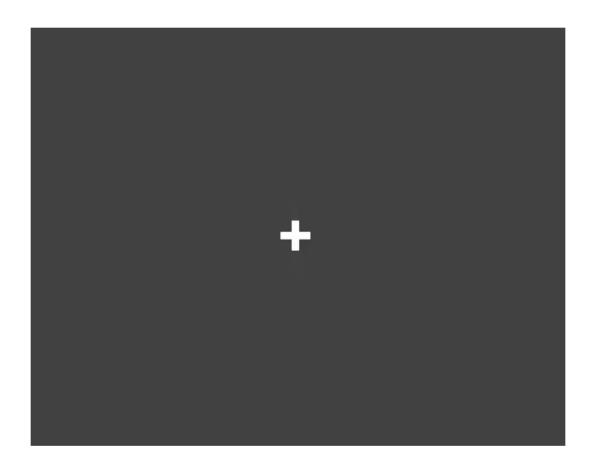


Conditions that make subjects look more similar to one another actually make better databases for identification:



# Q. Is rest best?A. Probably not

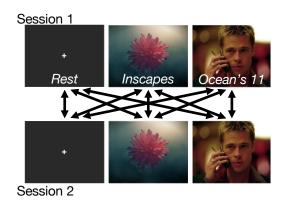
#### Consider naturalistic tasks:

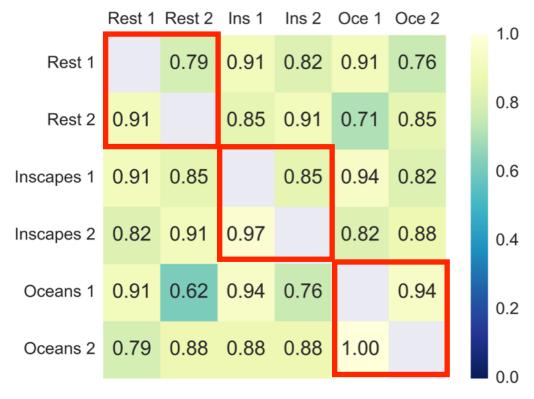


## Q. Is rest best? A. Probably not

#### Consider naturalistic tasks:

ID rate is just as good as (if not better than) rest

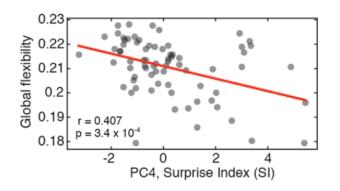


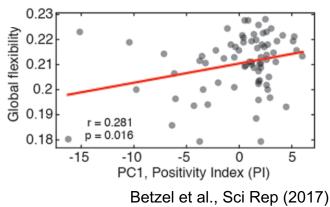


## How to choose behavior

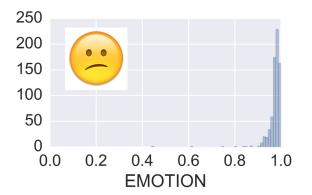
#### Is it stable?

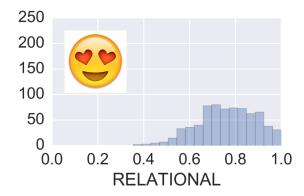
- Trait vs. state
- State variables may be better suited to within-subject analysis





Does it show a good distribution in your population?





# Behavior: Mitigating confounds

#### Many behaviors/phenotypes are correlated with head motion!

#### Negatively:

#### Positively:

		,	
Subject measures	Pearson r		
ReadEng (AgeAdj)	-0.23	DSM somatic problems (pct)	0.16
ReadEng (Unadj)	-0.23	DSM antisocial (raw)	0.16
Vocabulary (AgeAdj)	-0.19	ASR externalizing (raw)	0.16
Dexterity (Unadj)	-0.18	DSM somatic problems (raw)	0.16
CardSort (Unadj)	-0.18	Tobacco use 7 day	0.18
Dexterity (AgeAdj)	-0.18	Diastolic blood pressure	0.18
CardSort (AgeAdj)	-0.18	ASR externalizing	0.18
Education	-0.17	Tobacco use today	0.2
Fluid intelligence	-0.17	Systolic blood pressure	0.23
Spatial orientation	-0.17	Weight	0.52
Vocabulary (unadjj)	-0.17	Body mass index (BMI)	0.66
Emotion recognition	-0.16		

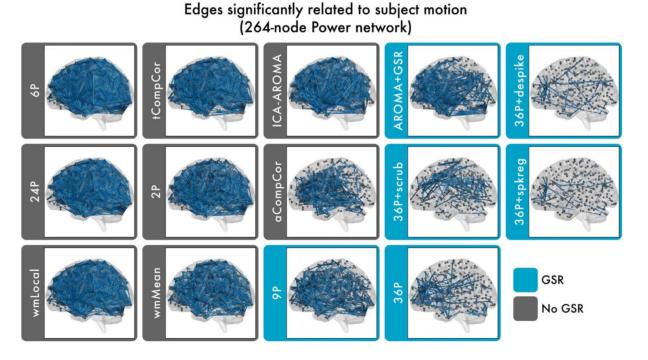
Siegel et al., Cerebral Cortex (2016)

- Patients of any kind move more
- Children move more
- Older adults move more

## Behavior: Mitigating confounds

Many behaviors/phenotypes are correlated with head motion!

- Check correlation in your sample
- Consider excluding particularly high-motion subjects
- Choose appropriate preprocessing techniques
- Use motion as an explicit covariate

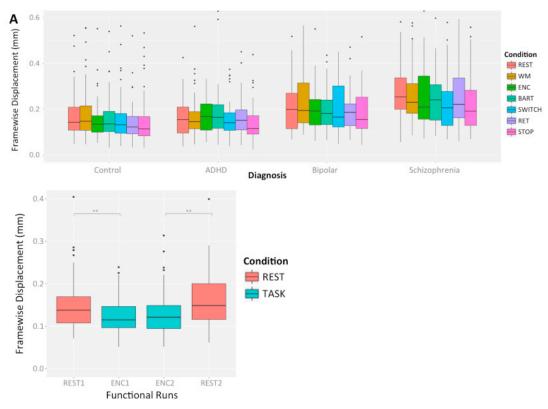


Ciric et al., Neurolmage (2017)

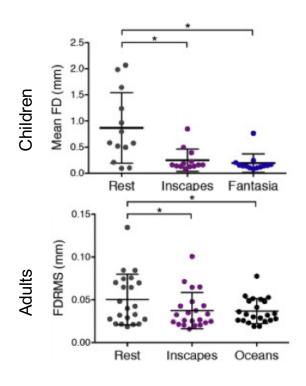
# Q. Is rest best? A. Probably not

#### Tasks also have purely practical advantages:

increase subject compliance (i.e., decrease head motion), especially in certain populations



Huijbers et al., Neurolmage (2017)

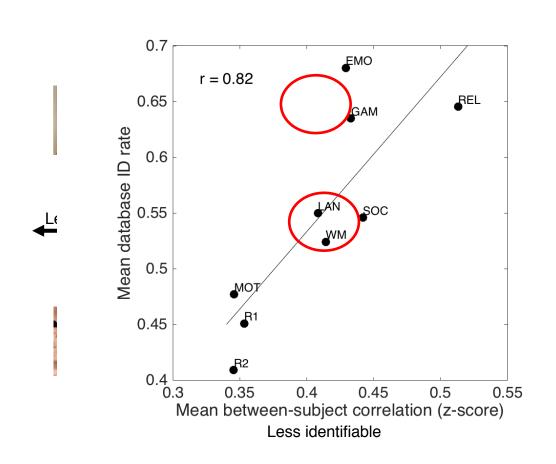


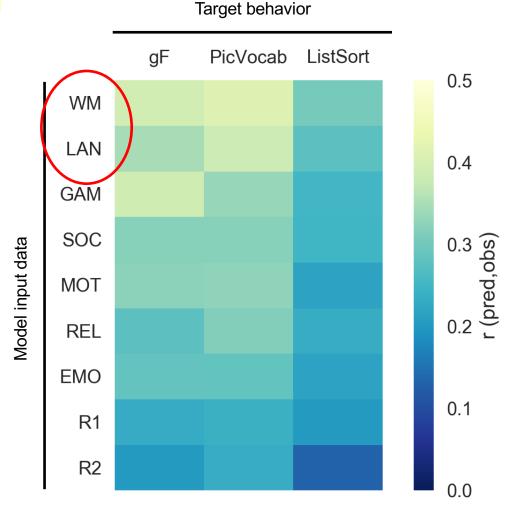
Vanderwal et al., Neurolmage (2015)

### Q. What is the best brain state?

# A. Maybe it depends on your behavior

Certain task conditions generate better predictions of behavior:





n = 716, 10-fold cross-validation Connectome-based Predictive Modeling (CPM; Shen et al., *Nat Protocols* 2017)

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# Outstanding questions

### Data acquisition

#### **Populations**

- Many individuals lightly sampled, or few individuals densely sampled?
- Patients, controls?
- Longitudinal studies?

#### **Imaging**

- Scan condition? "Stress test"?
- Function, anatomy, both?

#### **Behavior**

- Robust measures?
- Inter- vs intra-subject variability?

### Data analysis

#### Strategy

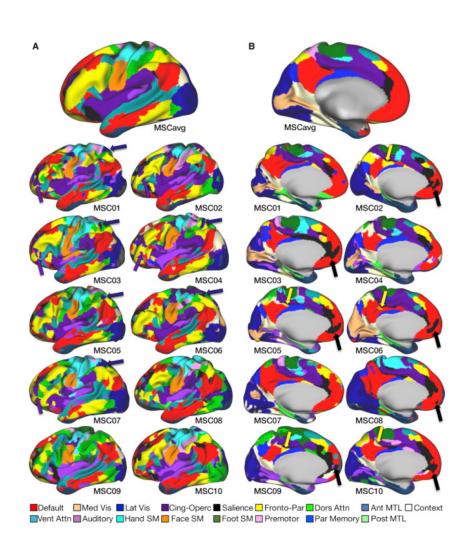
- Functional connectivity?
- Activation?
- Combination?

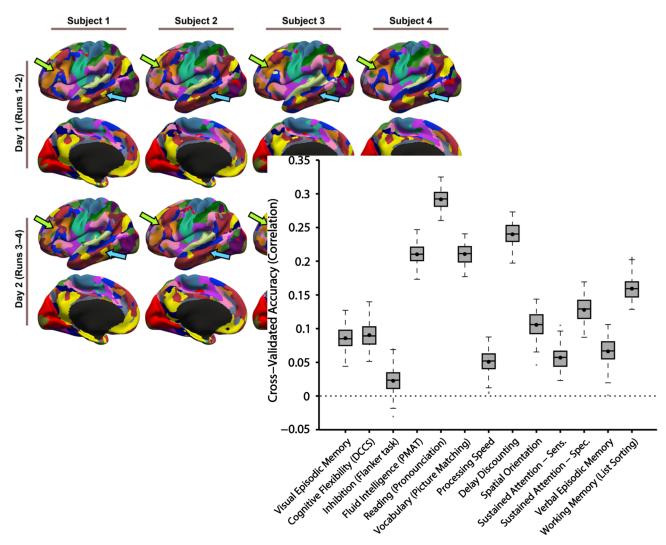
#### **Specifics**

 Parcel boundaries?
 Connections between parcels? Both?

# Interpretations & applications

# Individual-specific parcellations





# Outstanding questions

### Data acquisition

#### **Populations**

- Many individuals lightly sampled, or few individuals densely sampled?
- Patients, controls?
- Longitudinal studies?

#### **Imaging**

- Multisite studies?
- Scan condition? "Stress test"?
- Function, anatomy, both?

#### **Behavior**

- Robust measures?
- Inter- vs intra-subject variability?

### Data analysis

#### Strategy

- Functional connectivity?
- Activation?
- Combination?

#### **Specifics**

 Parcel boundaries?
 Connections between parcels? Both?

# Applications & interpretations

#### Mutability

- Development?
- Disease progression?
- Plasticity/training?

#### **Applications**

- Translational utility?
- Ethics?

# Further reading & open data sets

Selected reviews:

Prediction as a humanitarian and pragmatic contribution from human cognitive neuroscience Gabrieli, Ghosh & Gabrieli, Neuron (2015)

Building a science of individual differences from fMRI Dubois & Adolphs, *Trends in Cognitive Sciences* (2016)

From regions to connections and networks: new bridges between brain and behavior

Misic & Sporns, Current Opinion in Neurobiology (2016)

Can brain state be manipulated to emphasize individual differences in functional connectivity? Finn et al., *NeuroImage* (2017)

Open data sets with brain and behavior:















Use these on their own or in combination with your own data to generate or test hypotheses, see if a finding generalizes, etc

# Acknowledgements



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Marvin Chun
Monica Rosenberg
Tamara Vanderwal



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