

Neurotopics: Unsupervised learning for discovering functional regions of the brain

Timothy Rubin, Michael N. Jones,
Oluwasanmi Koyejo, Tal Yarkoni

Introduction

- A primary goal of cognitive neuroscience: identifying relationships between the brain and cognitive functions
- Traditional fMRI studies focus on identifying individual functional regions
 - Small sample sizes, low statistical power
- More recently: Big-data approaches based on statistical learning of corpus data

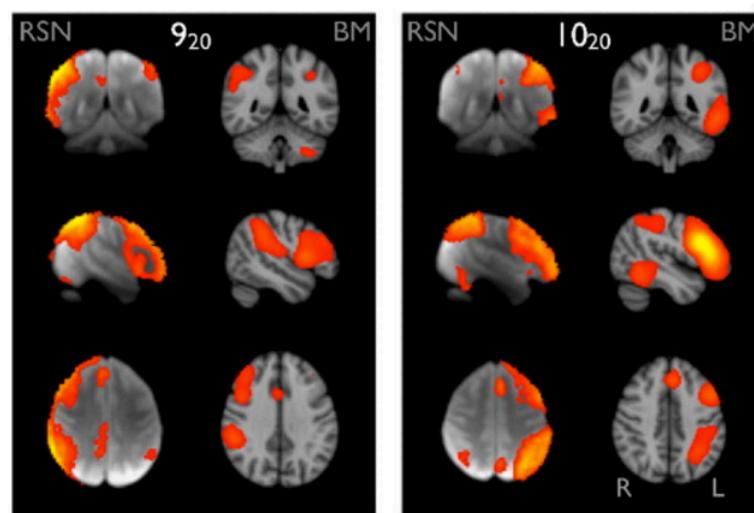
Statistical learning approaches

Two primary approaches:

Statistical learning approaches

Two primary approaches:

1. Network mappings of functional activation using dimensionality reduction (e.g., ICA or PCA)
 - These networks are spatially diffuse
 - Span multiple regions with distinct functions



Statistical learning approaches

Two primary approaches:

1. Network mappings of functional activation using dimensionality reduction (e.g., ICA or PCA)
 - These networks are spatially diffuse
 - Span multiple regions with distinct functions
2. Clustering/parcellation type approaches (e.g. k-means clustering)
 - Hard-assignments of all voxels to a single region
 - Does not allow for single location to be involved in multiple functions (i.e., no “one-to-many” mapping)

Statistical learning approaches

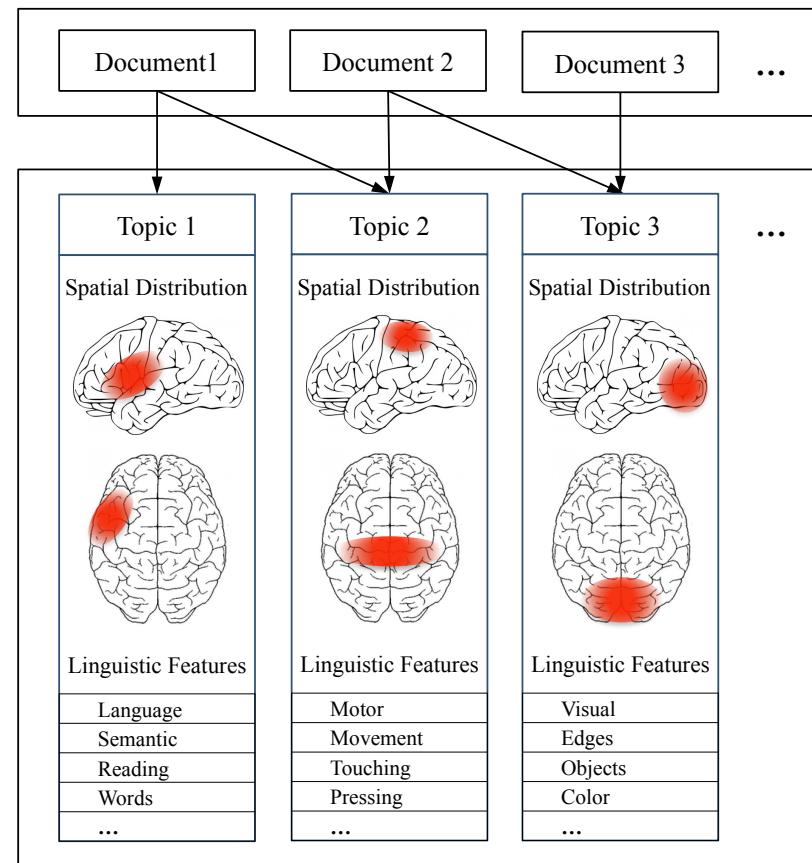
- Additional concern: spatial components are typically extracted *independent of function*:
 - First, spatial components are extracted
 - After, these components are mapped onto cognitive functions
- To optimally produce a mapping: these components should be jointly extracted

Goals of our approach

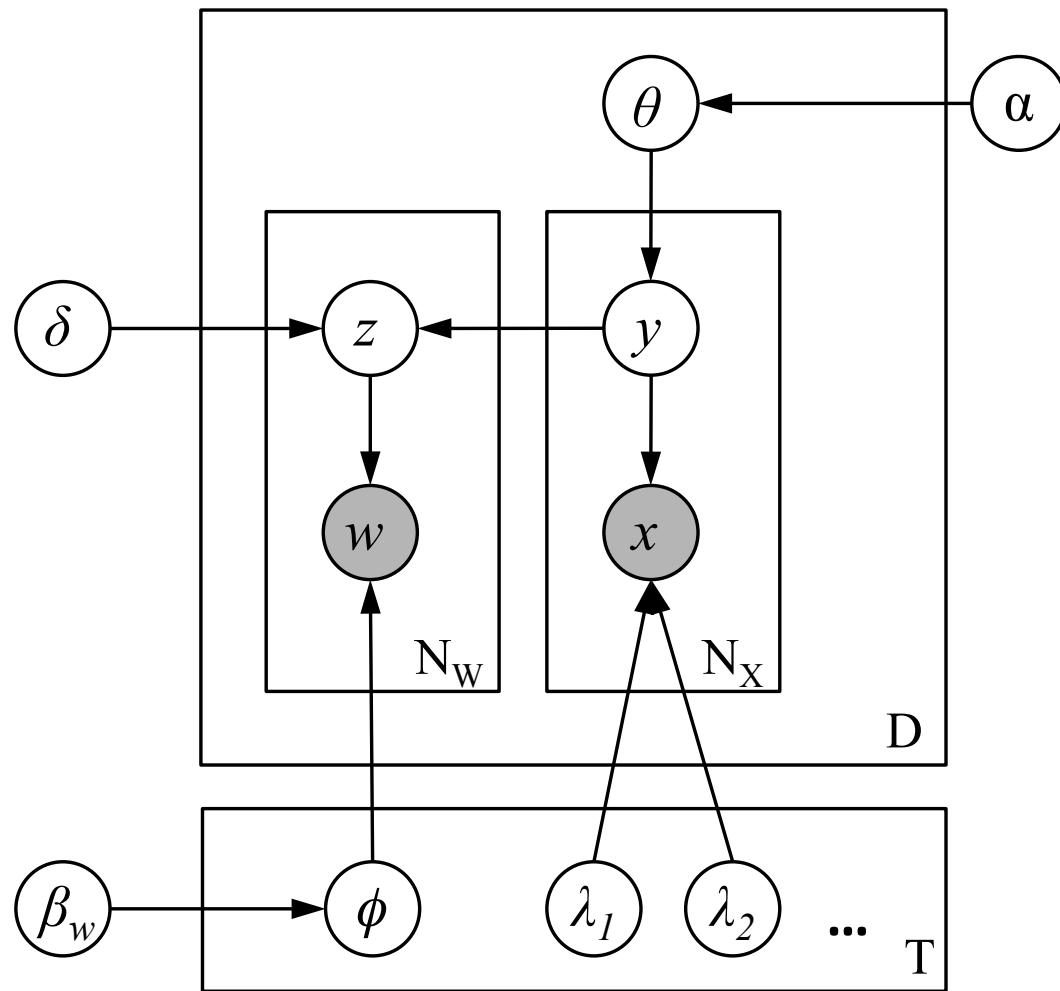
- Identify functional units of brain by *jointly* modeling their spatial and functional components using corpus data
- Allow for “Soft clustering”: a single location can belong to multiple functional regions
- Sensitive to both anatomical and psychological constraints
 - e.g. lateral symmetry, localized function

Generative topic model

- Based on Correspondence LDA model
- Broad idea:
 - Each document is modeled as a mixture of *topics* (i.e., regions)
 - Each region consists of:
 - A spatial probability distribution over neural *activations*
 - A linguistic probability distribution over terms related to the regions function



Graphical model



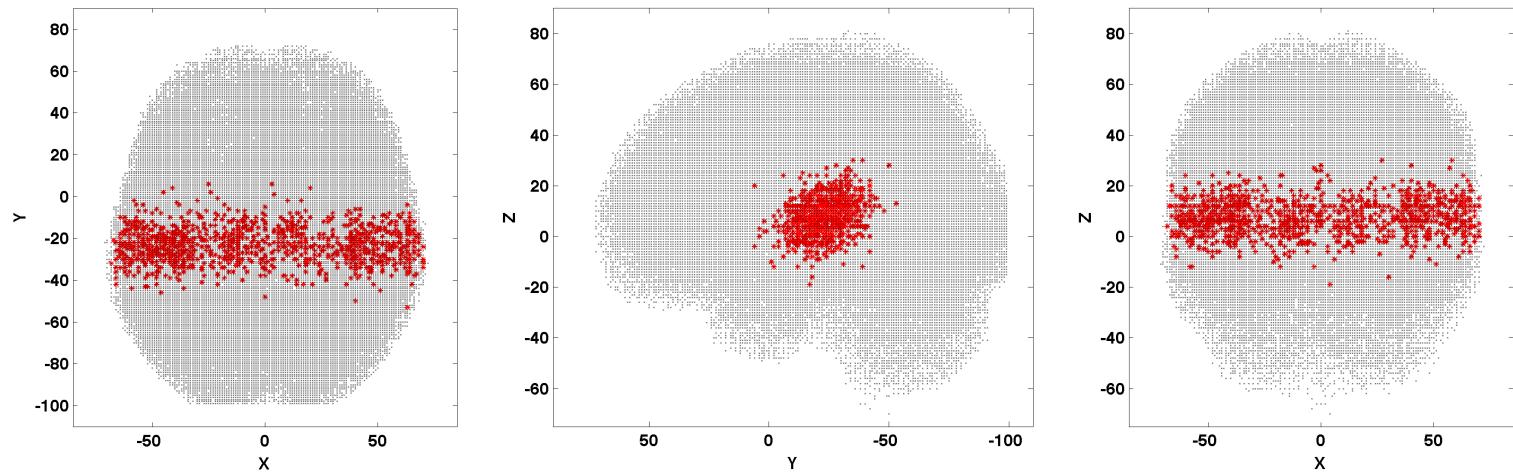
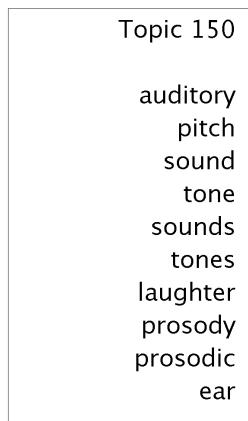
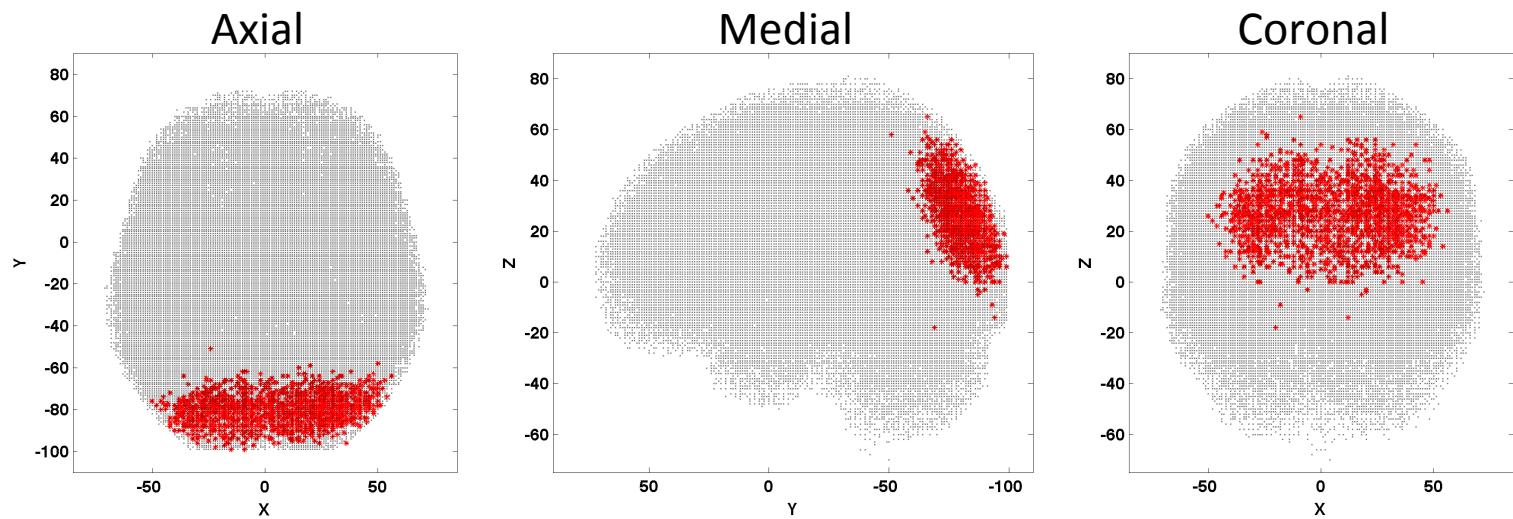
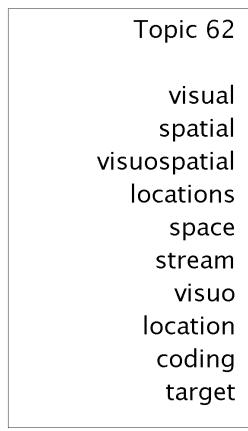
Training the model

- Trained on the Neurosynth corpus
 - About 11,000 fMRI publications
 - Text from abstracts (~500k tokens)
 - Reported (x,y,z) peak activation coordinates (~400k tokens)
- Extracted 200 Topics
- Model is trained using Bayesian methods

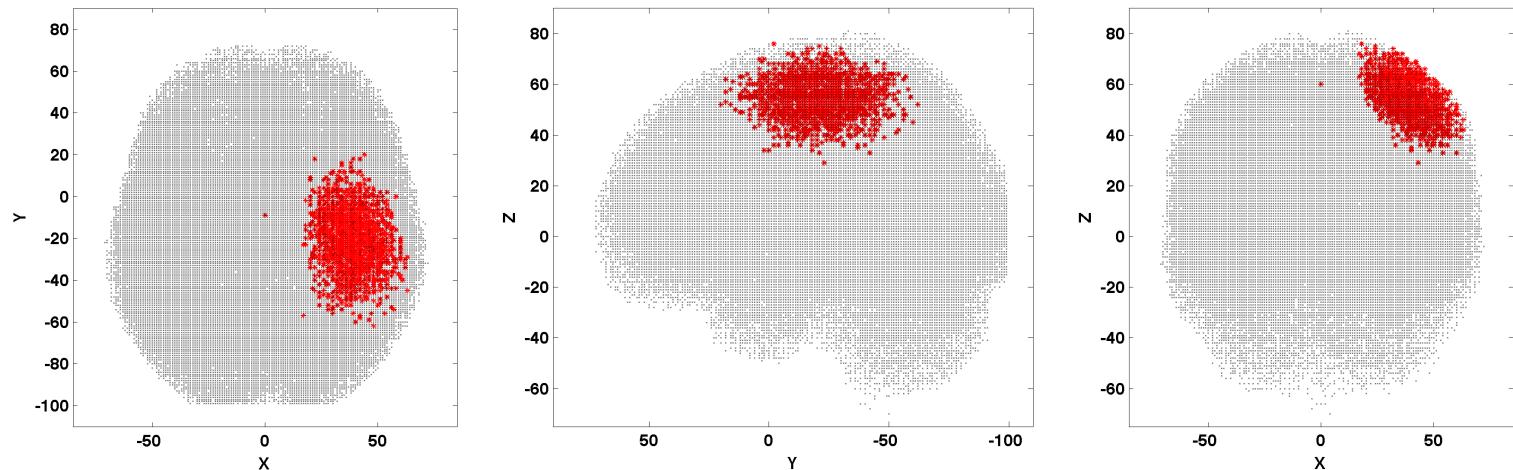
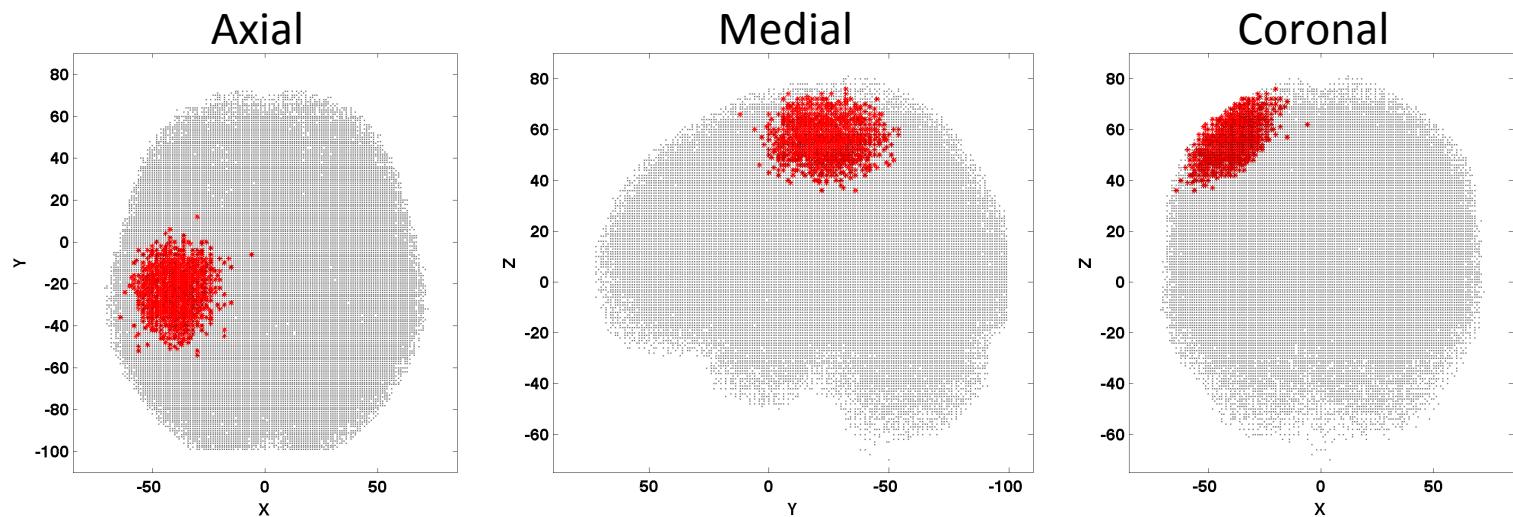
Regarding Spatial distributions:

- Our generalized version of correspondence-LDA allows the experimenter to choose an appropriate spatial distribution
- **Simple model:**
 - Regions spatial distribution represented by a single Gaussian distribution

Results: Single-region model



Results: Single-region model



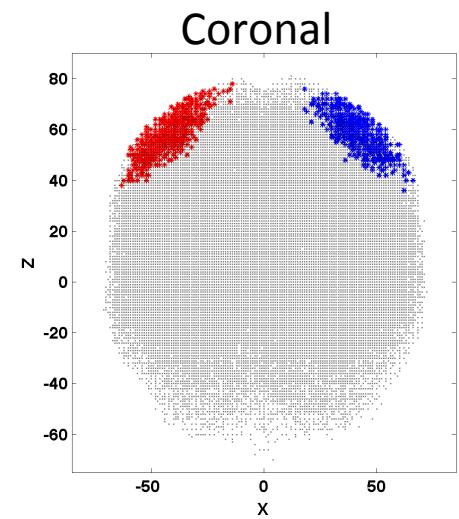
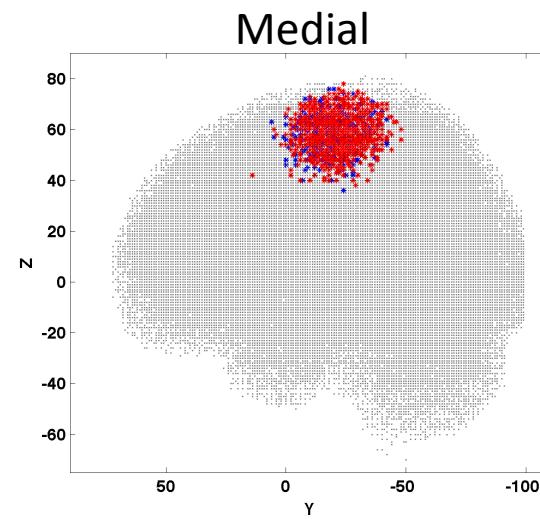
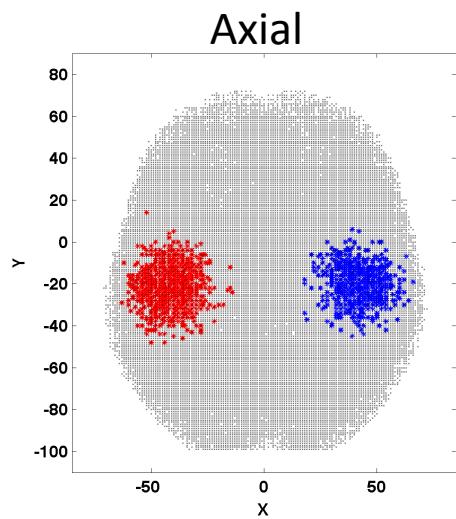
Spatial distributions:

- **Anatomically constrained model:**
 - Regions represented by a weighted mixture of two Gaussian distributions
 - Laterally symmetric (with respect to the brain hemispheres)

Results: Symmetric Subregions

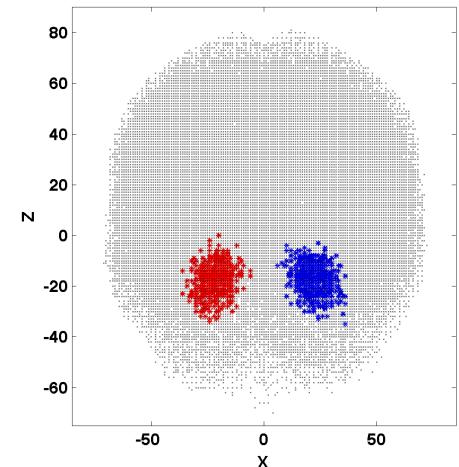
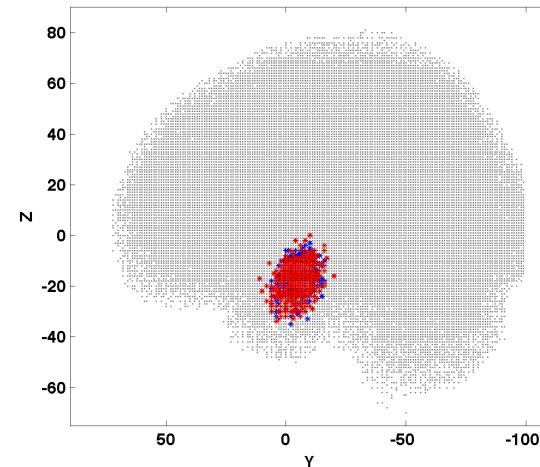
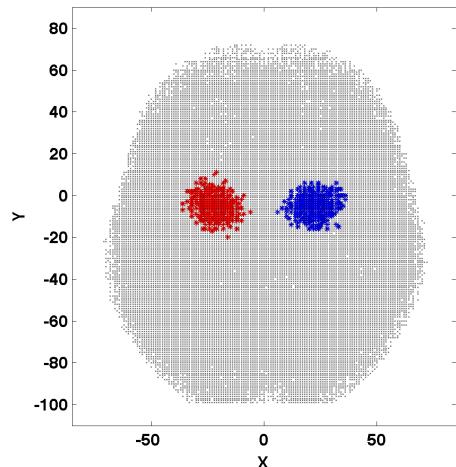
Topic 79

motor
stimulation
finger
somatosensory
transcranial
sensorimotor
repetitive
mep
plasticity
handed



Topic 29

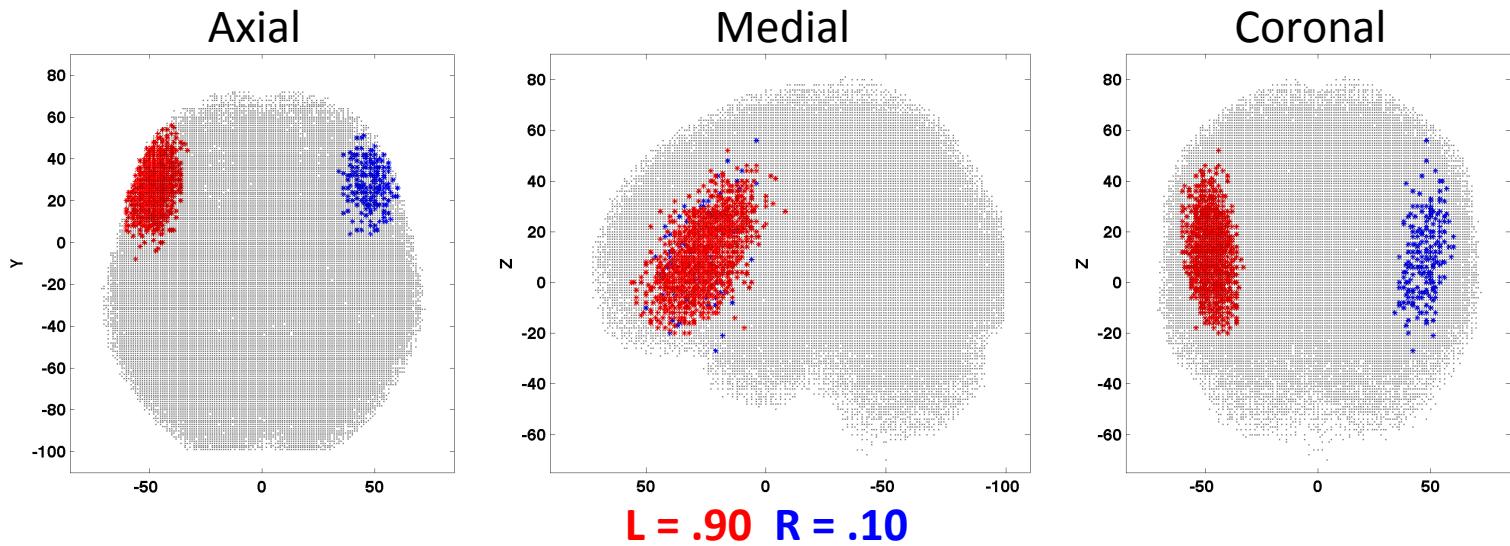
emotional
emotion
pictures
affective
faces
ratings
affect
fearful
perception
distraction



Functional Lateralization

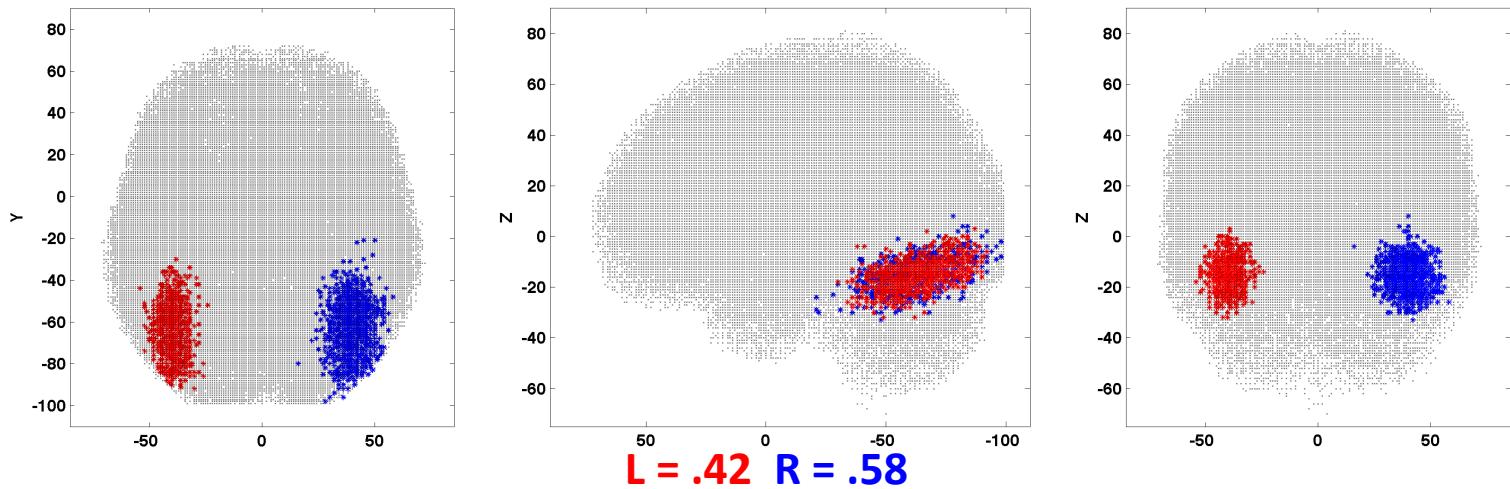
Topic 200

- semantic fluency
- semantic_processing selection
- retrieval verbal demands
- judgment meaning
- verbal_fluency



Topic 09

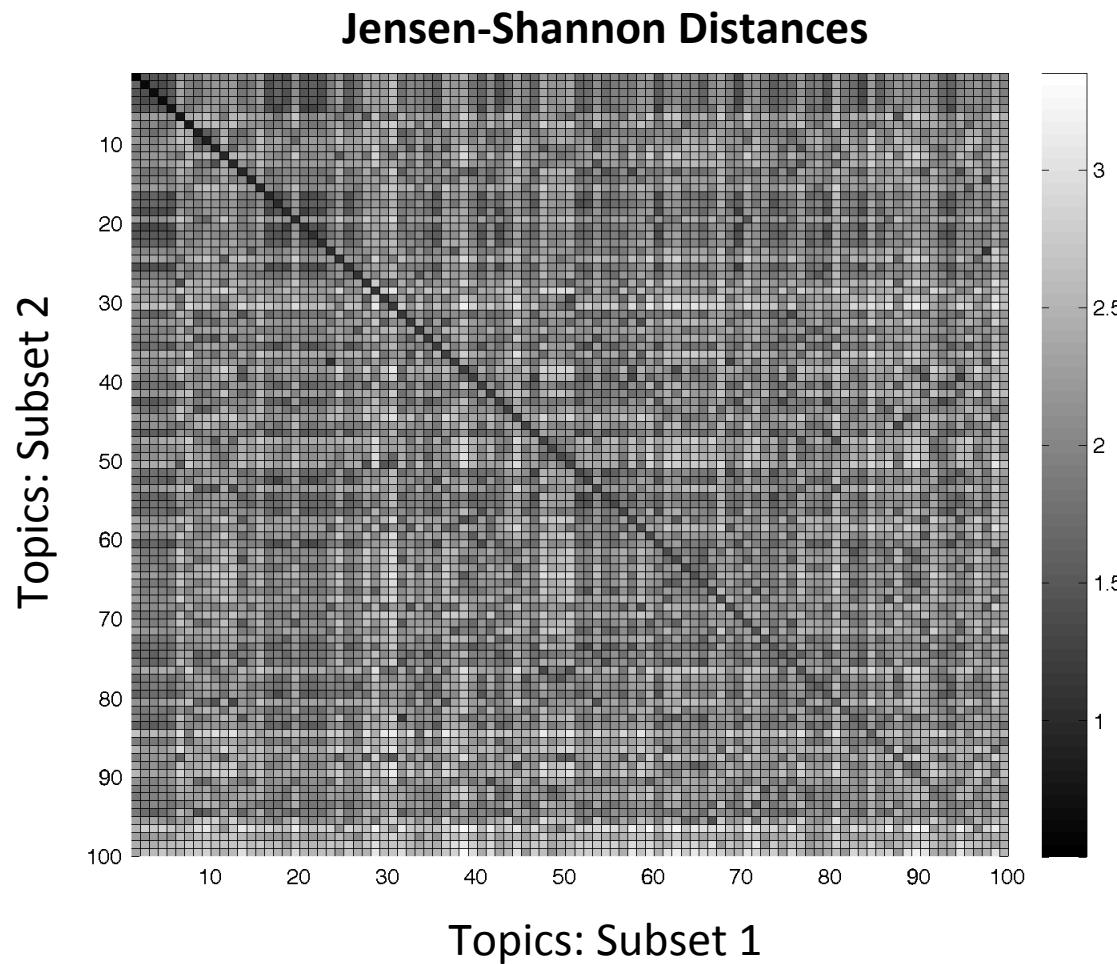
- face
- faces
- identity
- facial
- face_recognition
- recognition
- face_perception
- selectivity
- upright
- inversion



Topic stability analysis

- Split the dataset in half and trained a symmetric-subregion model separately on each half
- Aligned the topics based on the similarities of the linguistic + spatial distributions
- Evaluate how many topics are consistent independent of the specific training data

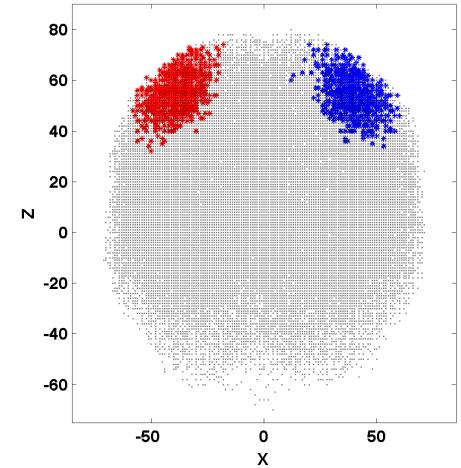
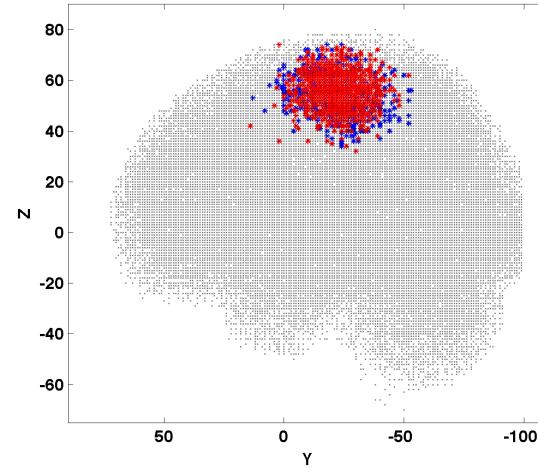
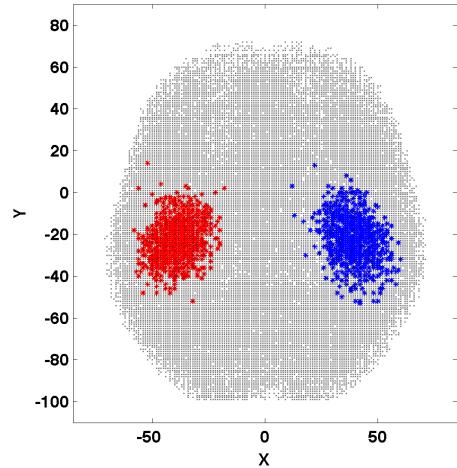
Topic-alignment across datasets



Rank 10

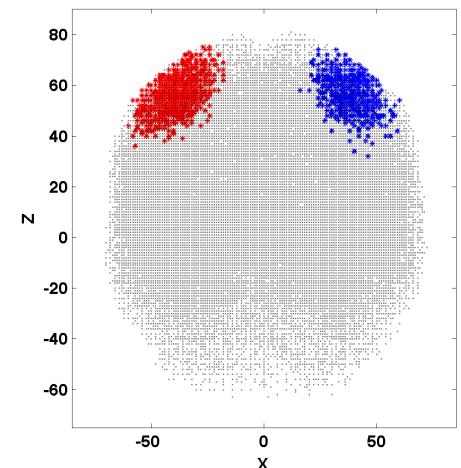
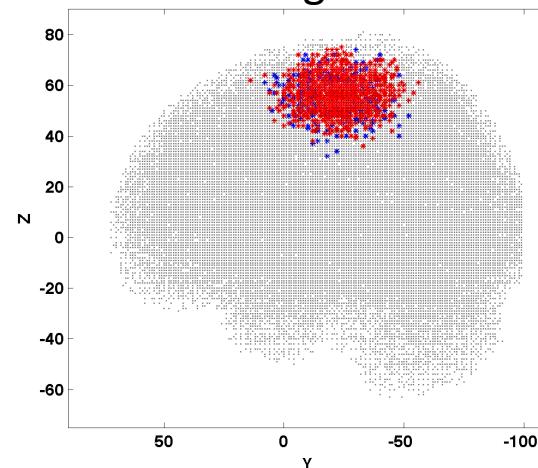
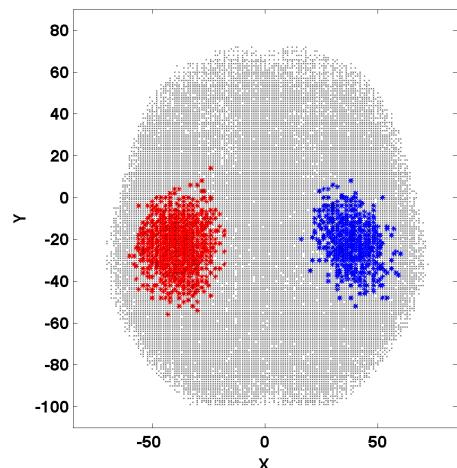
Topic 55

- motor movement finger
- somatosensory force
- movements sensorimotor stimulation
- grip fingers



Topic 51

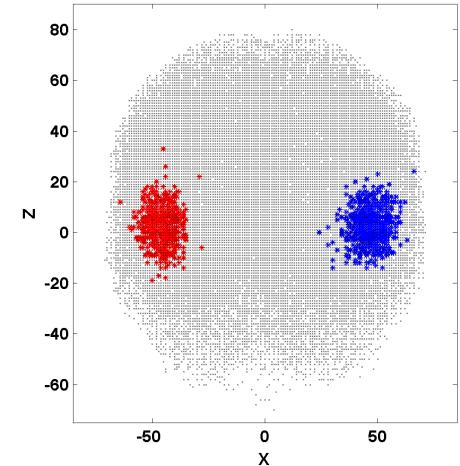
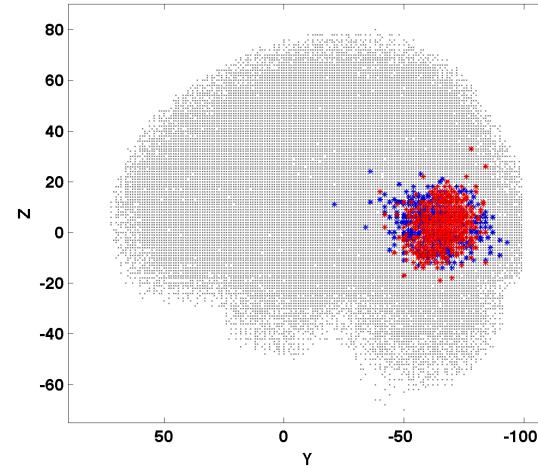
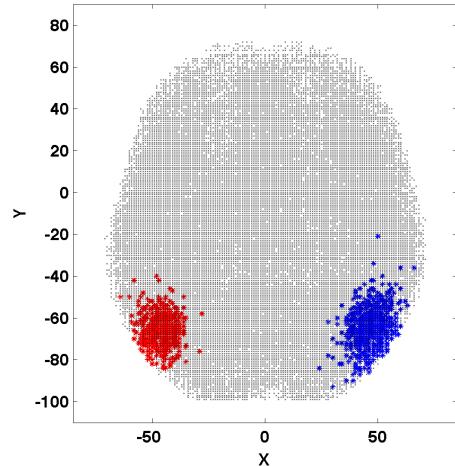
- motor finger
- movements stimulation
- bimanual sensorimotor movement
- alpha handed dominant



Rank 30

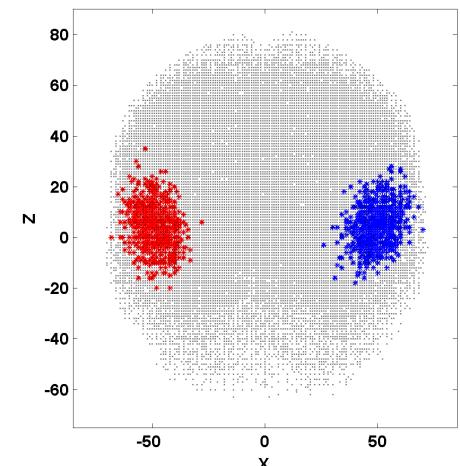
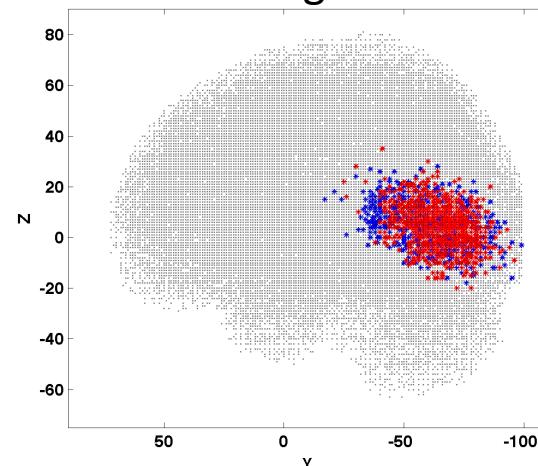
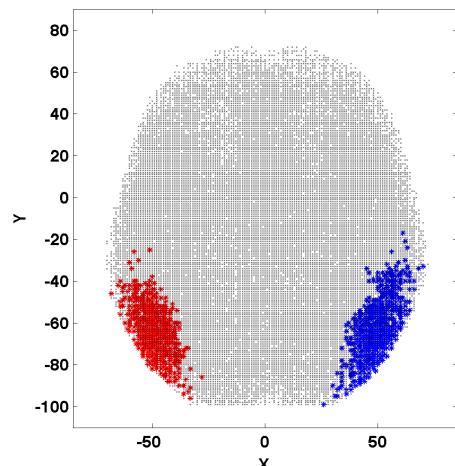
Topic 50

- motion
- visual
- moving
- static
- direction
- body
- biological_motion
- perception
- motion_perception
- flow



Topic 88

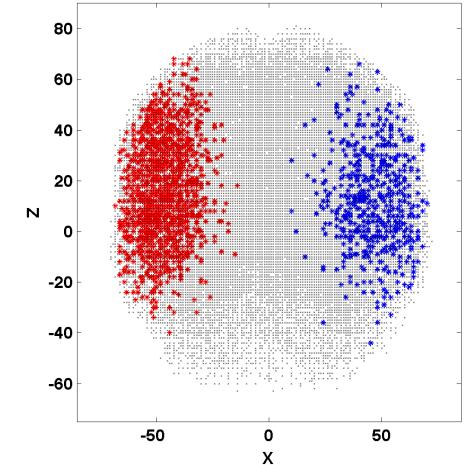
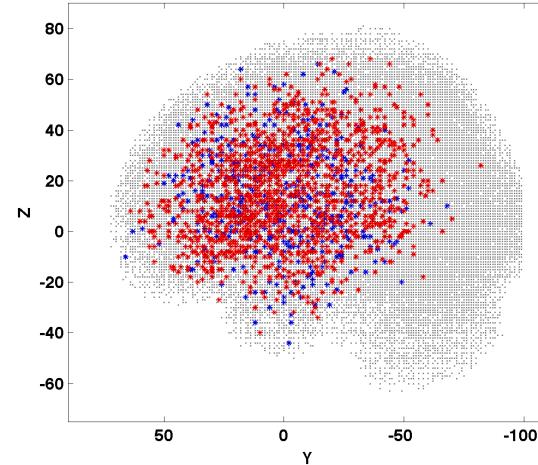
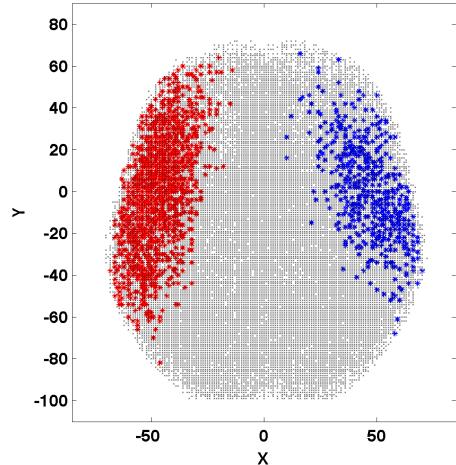
- body
- motion
- perception
- viewing
- actions
- biological_motion
- videos
- gestures
- viewed
- visual



Rank 50

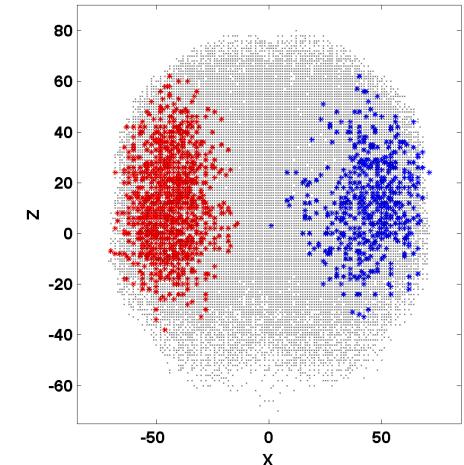
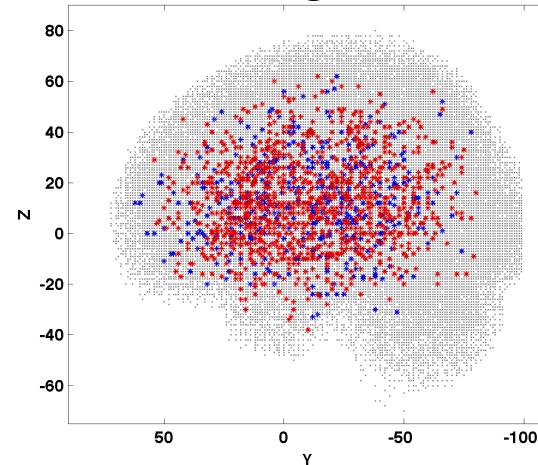
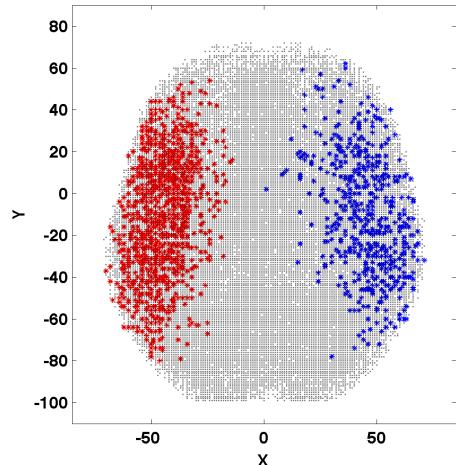
Topic 65

words
semantic
word
lexical
meaning
semantic_processing
picture
competition
language
semantically



Topic 84

language
native
chinese
languages
hearing
bilinguals
linguistic
japanese
deaf
speakers

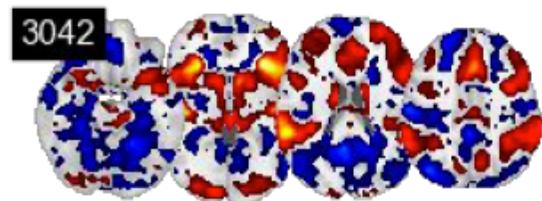


Stability analysis

- Approximately 50% of topics are stable, independent of the training data

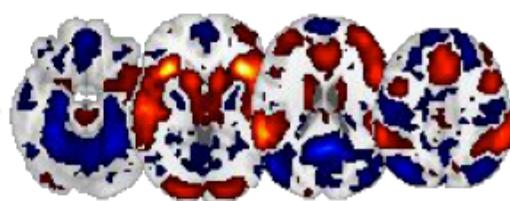
Image Decoding

Original map



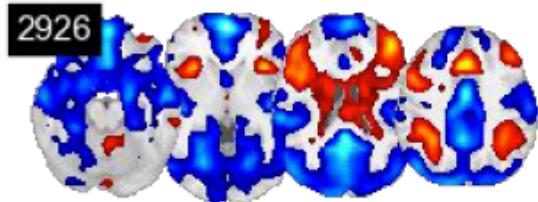
$R^2 = 0.70$

Reconstructed map

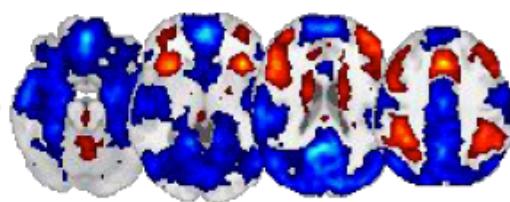


→

music inhibition
speech
incongruent
prosody
auditory
cues
voice
verbal control
stimulus conflict
language
kappa

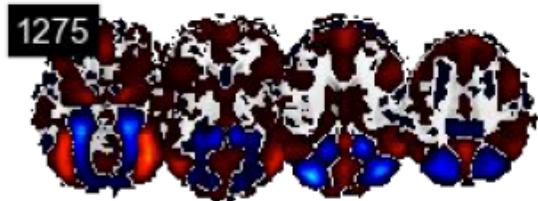


$R^2 = 0.70$

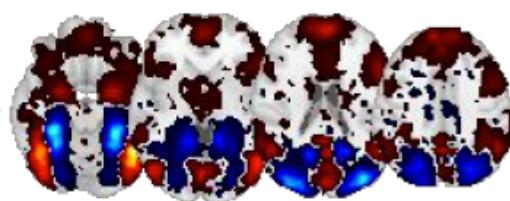


→

memory task
single trials
inhibition language
task selection time
switching WM task error cues
conflict control number
attention activation



$R^2 = 0.73$



→

connectivity depression disgust anxiety patient subjects
moral negative moral social
emotional emotional emotion
emotional judgments empathy mental personality
schizophrenia social network brain held activation

Discussion

- Large scale statistical learning approach that learns interpretable functional regions
- Regions are consistent with the literature
- Addresses drawbacks of previous approaches (e.g., “one-to-many” mapping)
- Additionally, provide a measure of the lateralization of cognitive functions

Thanks