

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

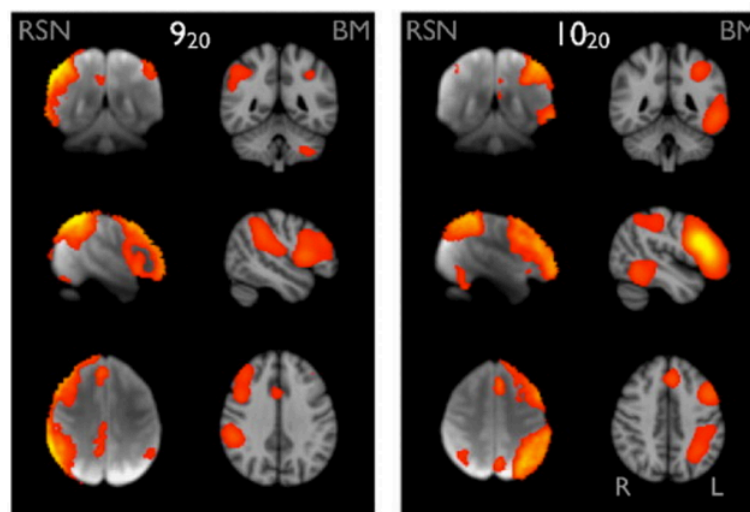
Statistical learning approaches

Two primary approaches:

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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



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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

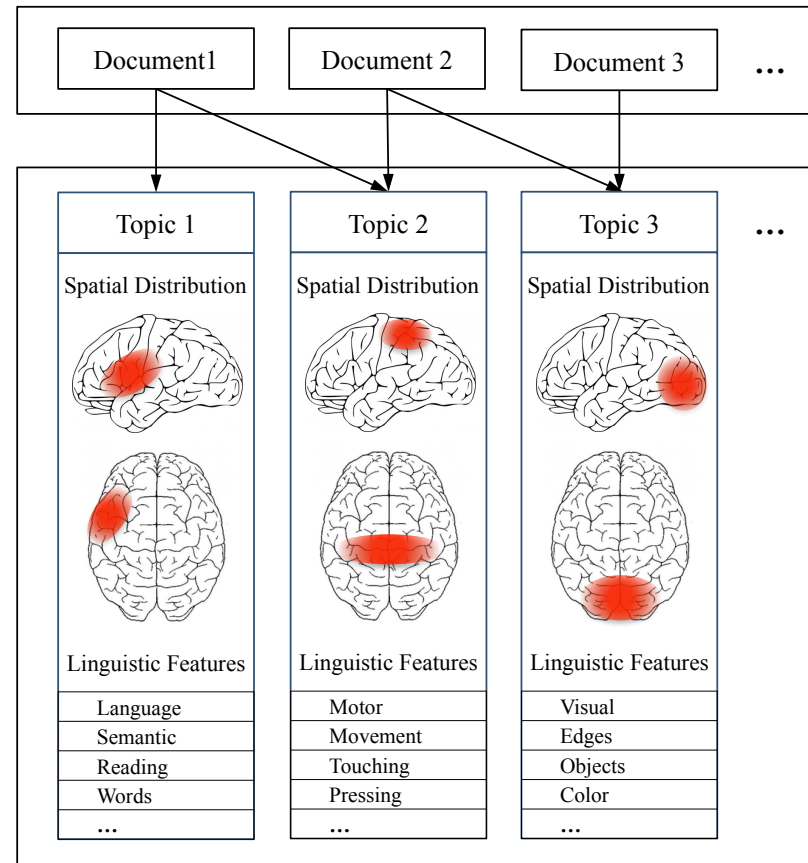
- Additional concern: spatial components are typically extracted *independent of function*:
 - First, spatial components are extracted
 - After, these components are mapped onto cognitive functions
- To optimally produce a mapping: these components should be jointly extracted

Goals of our approach

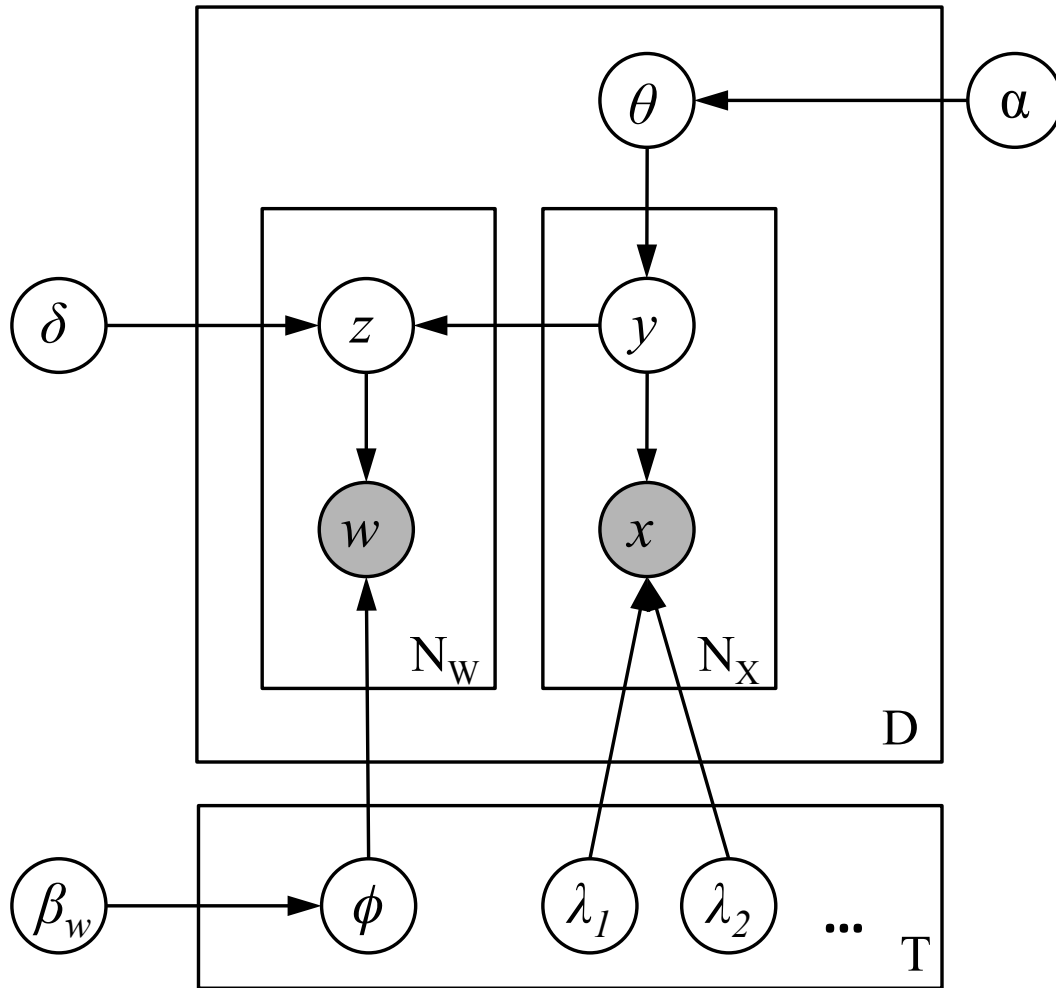
- Identify functional units of brain by *jointly* modeling their spatial and functional components using corpus data
- Allow for “Soft clustering”: a single location can belong to multiple functional regions
- Sensitive to both anatomical and psychological constraints
 - e.g. lateral symmetry, localized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

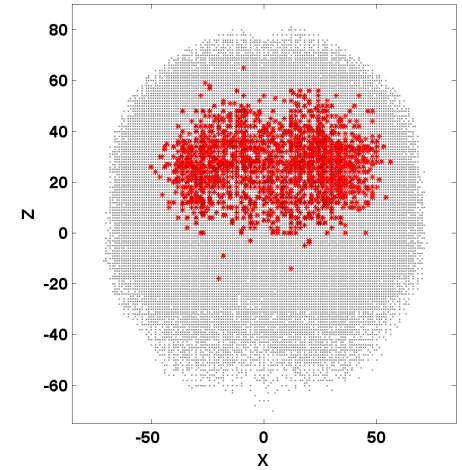
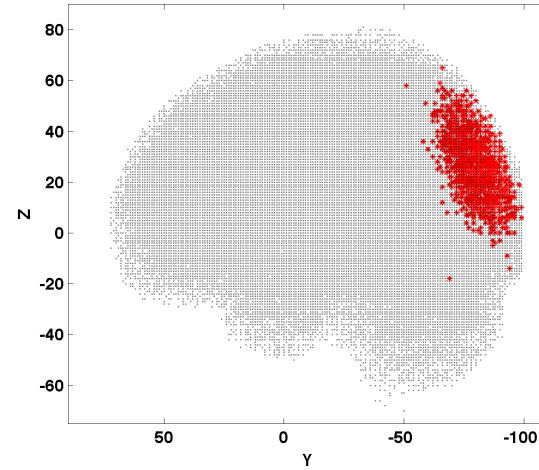
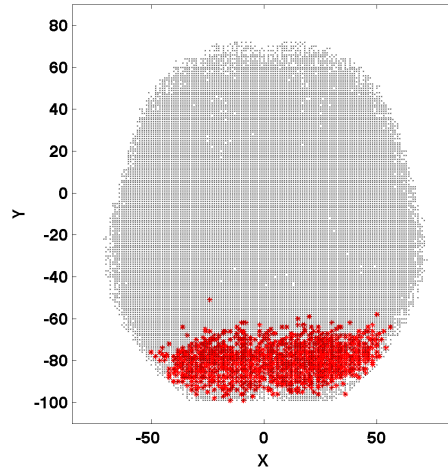
Axial

Medial

Coronal

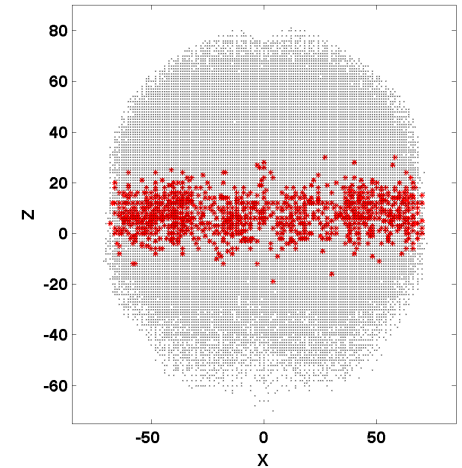
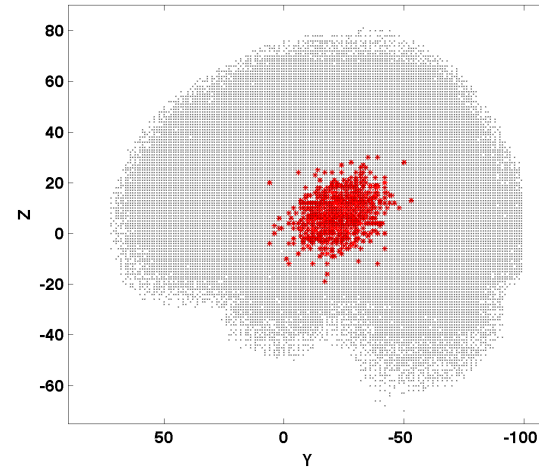
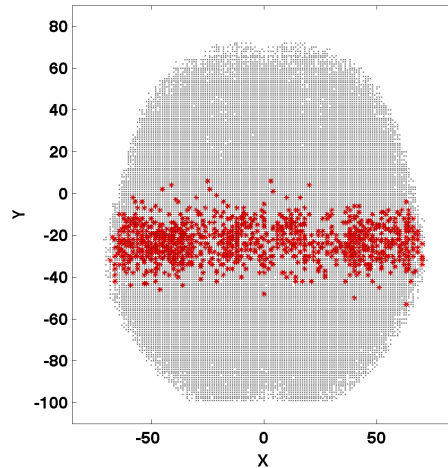
Topic 62

visual
spatial
visuospatial
locations
space
stream
visuo
location
coding
target



Topic 150

auditory
pitch
sound
tone
sounds
tones
laughter
prosody
prosodic
ear



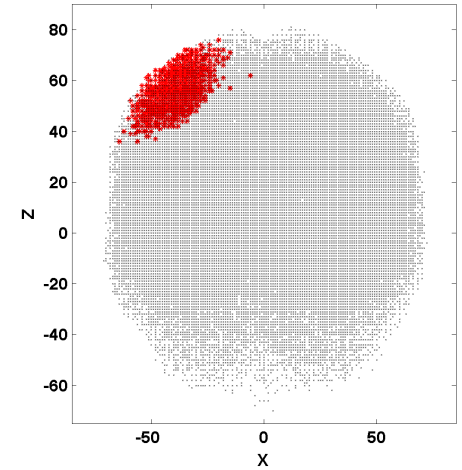
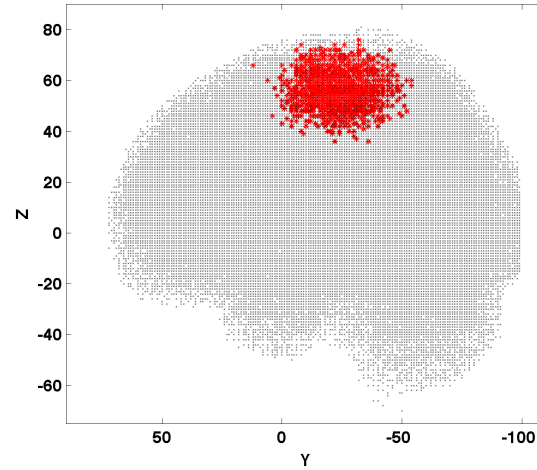
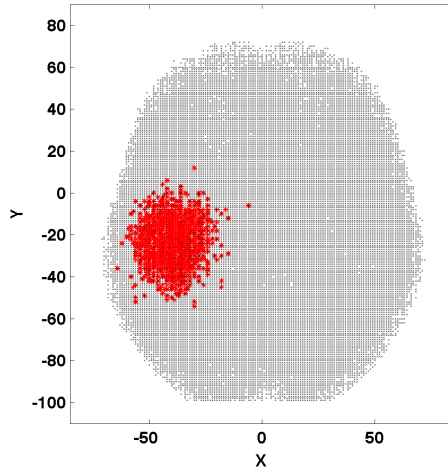
Results: Single-region model

Axial

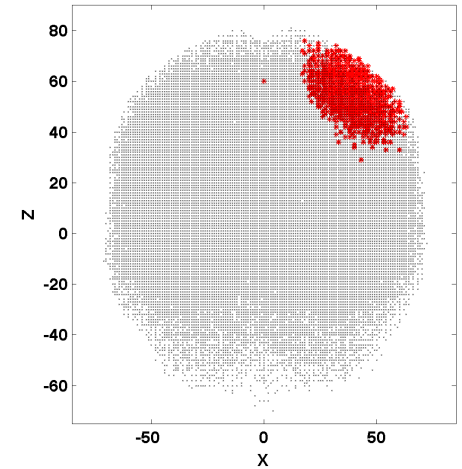
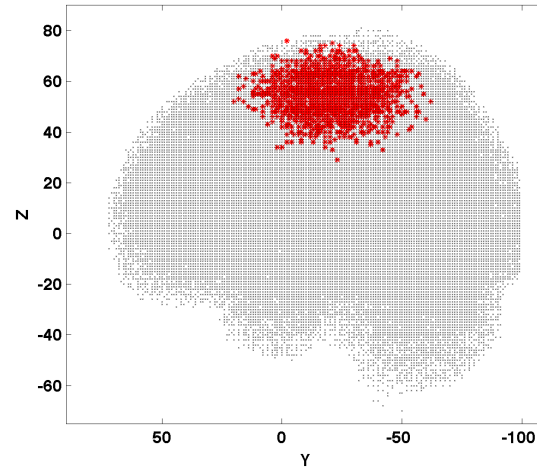
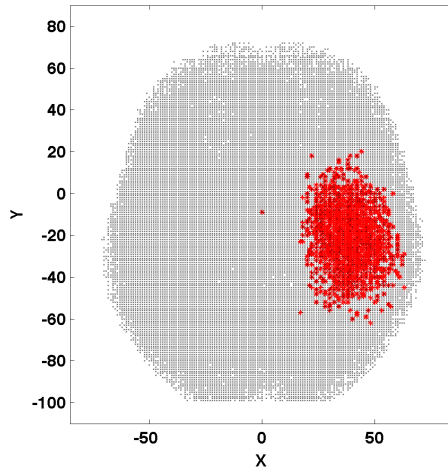
Medial

Coronal

Topic 117
motor
finger
changes
bimanual
somatosensory
sensorimotor
movements
movement
coordination
unimanual



Topic 152
motor
movement
movements
finger
changes
execution
sensorimotor
paced
handed
spatial

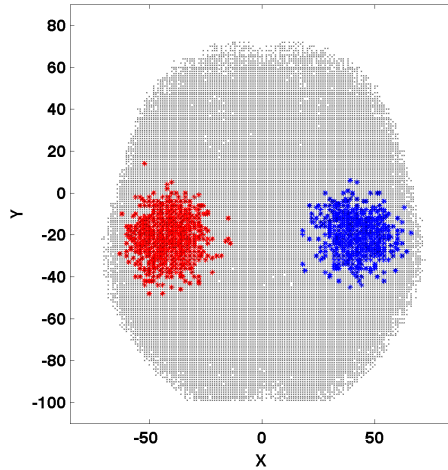


Spatial distributions:

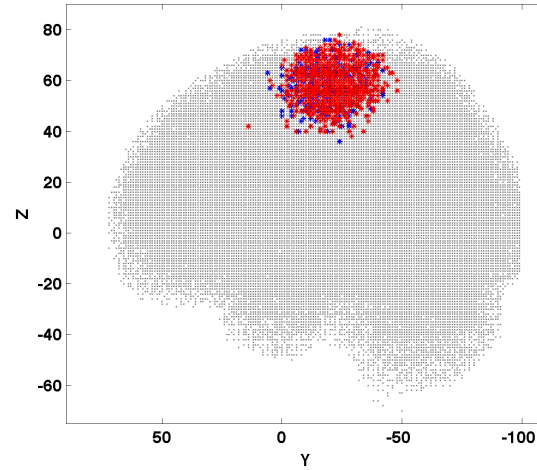
- **Anatomically constrained model:**
 - Regions represented by a weighted mixture of two Gaussian distributions
 - Laterally symmetric (with respect to the brain hemispheres)

Results: Symmetric Subregions

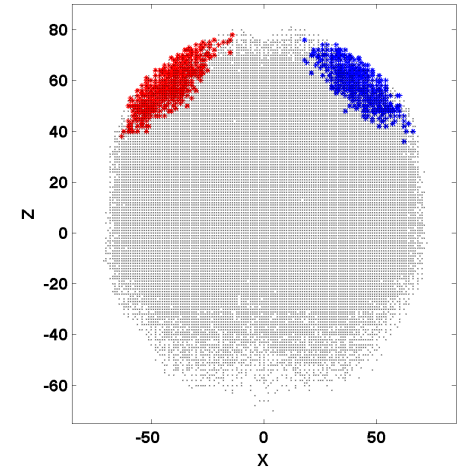
Axial



Medial



Coronal

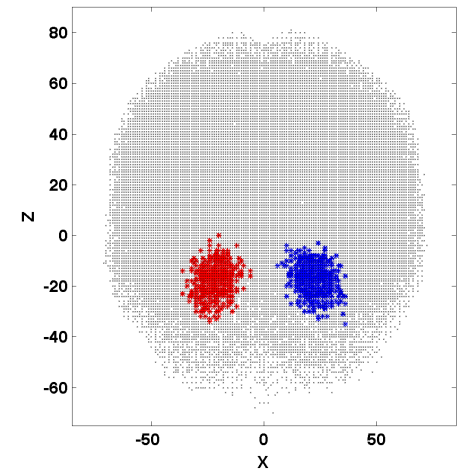
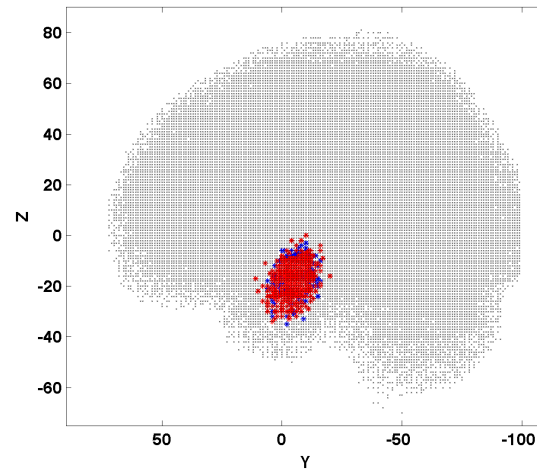
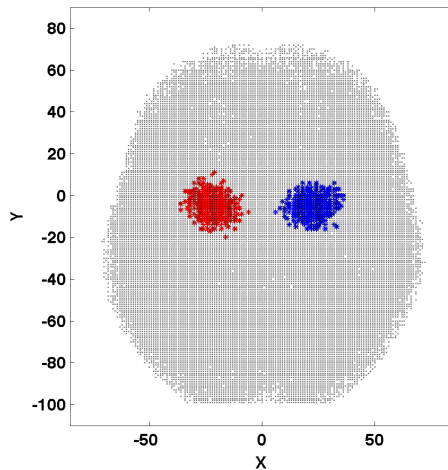


Topic 79

motor
stimulation
finger
somatosensory
transcranial
sensorimotor
repetitive
mep
plasticity
handed

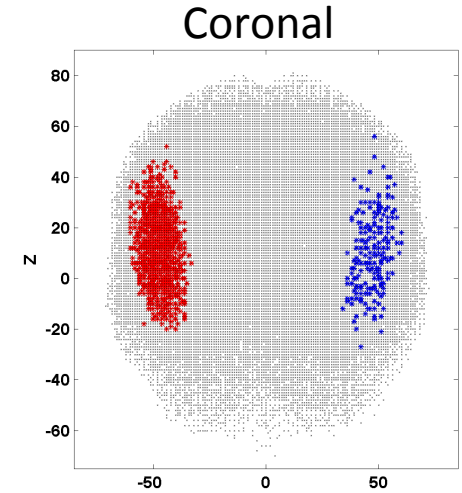
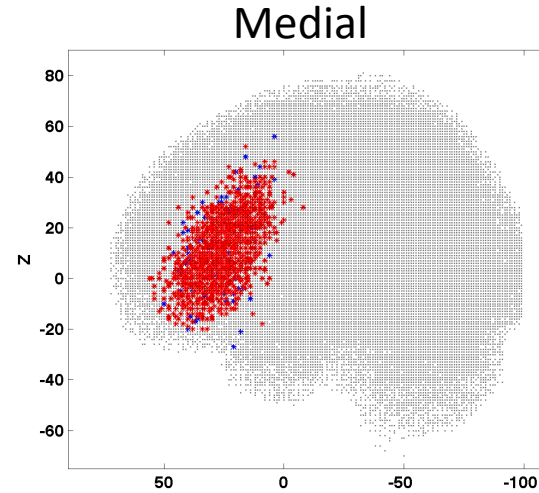
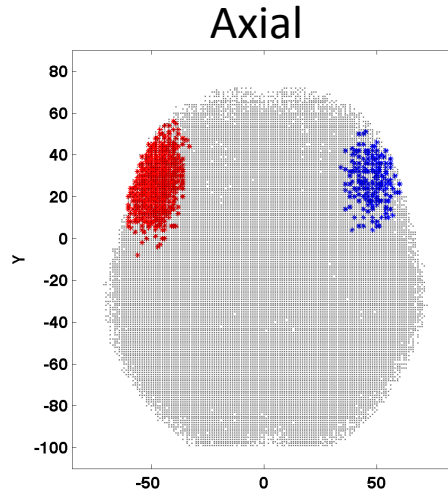
Topic 29

emotional
emotion
pictures
affective
faces
ratings
affect
fearful
perception
distraction



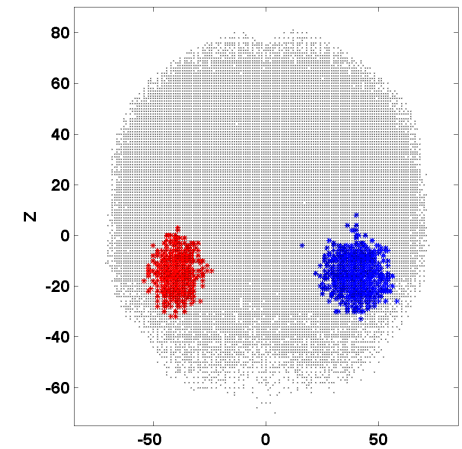
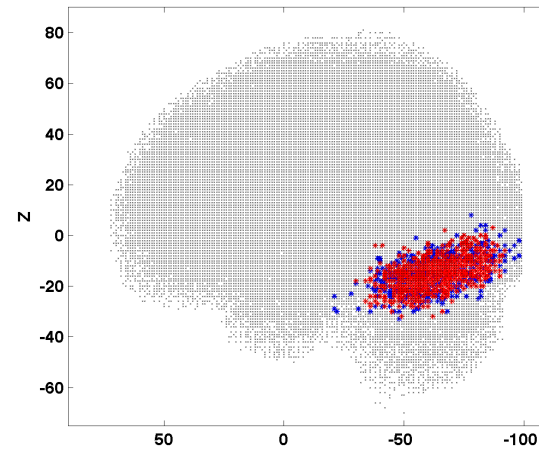
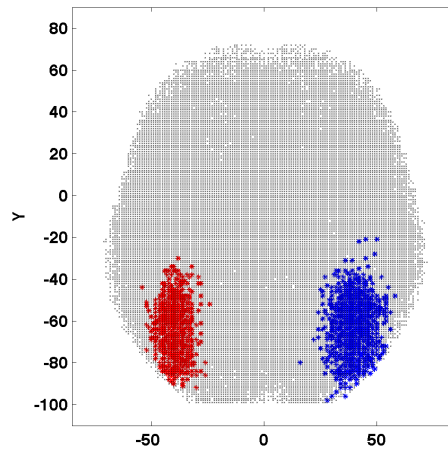
Functional Lateralization

Topic 200
semantic
fluency
semantic_processing
selection
retrieval
verbal
demands
judgment
meaning
verbal_fluency



L = .90 R = .10

Topic 09
face
faces
identity
facial
face_recognition
recognition
face_perception
selectivity
upright
inversion

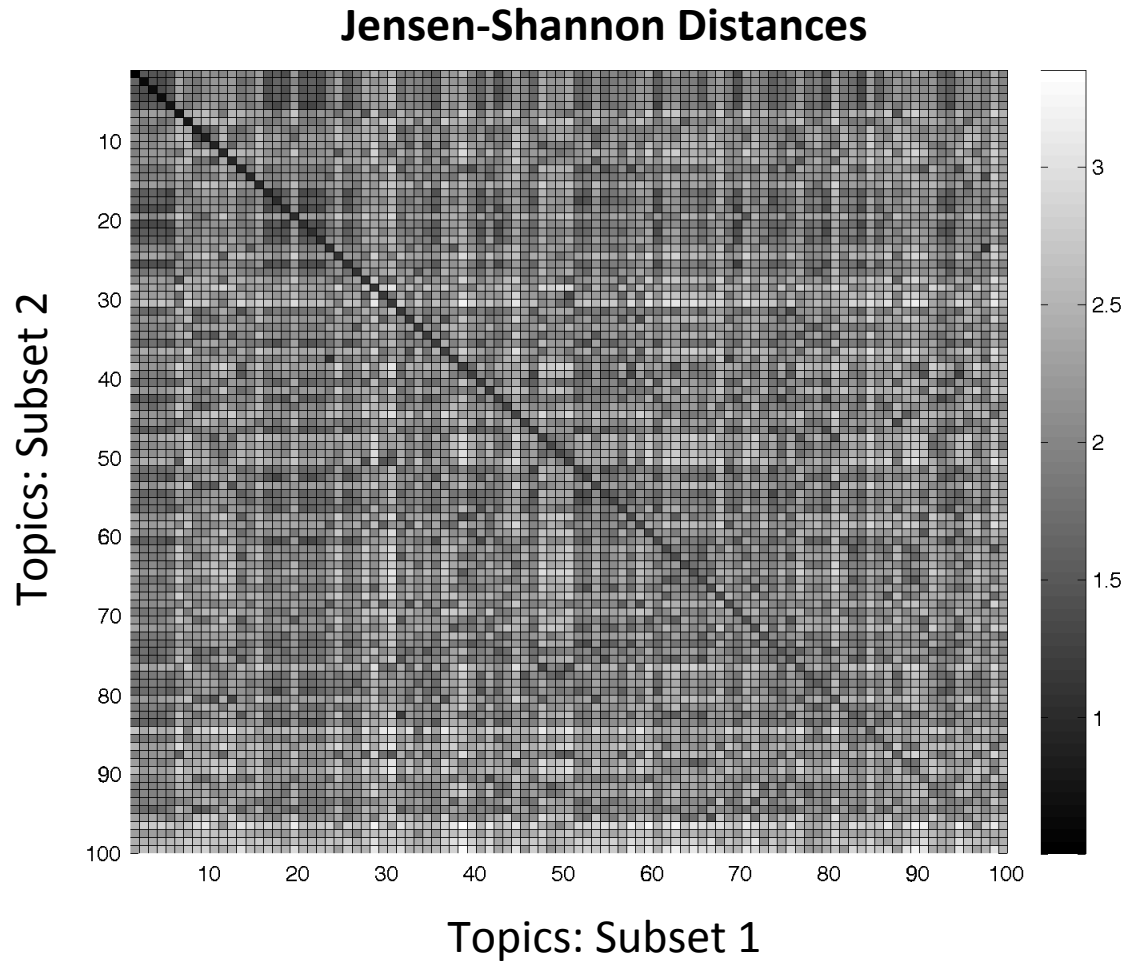


L = .42 R = .58

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

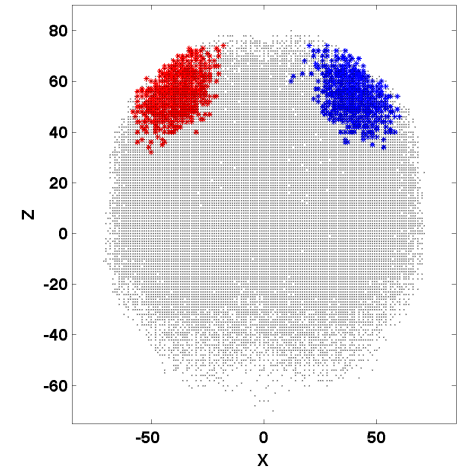
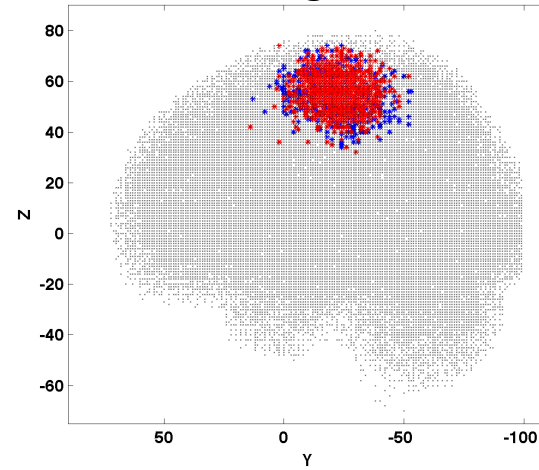
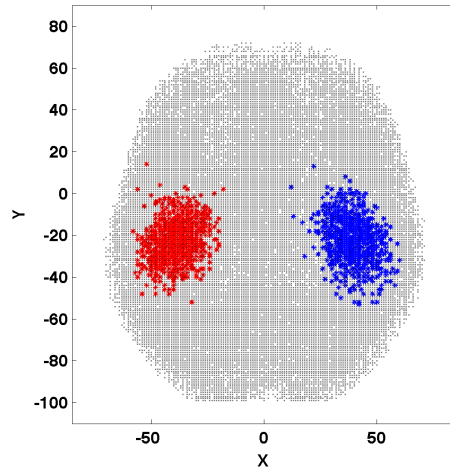
Topic-alignment across datasets



Rank 10

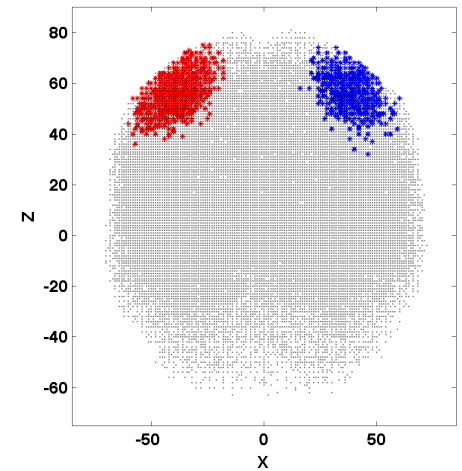
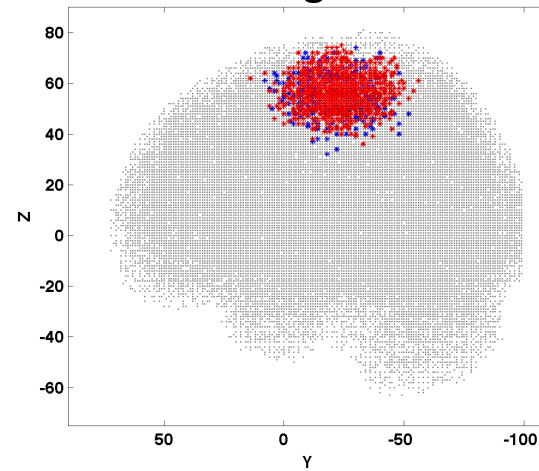
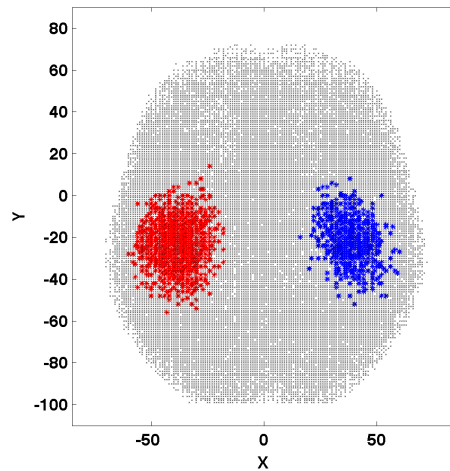
Training Set 1

Topic 55
motor
movement
finger
somatosensory
force
movements
sensorimotor
stimulation
grip
fingers



Training Set 2

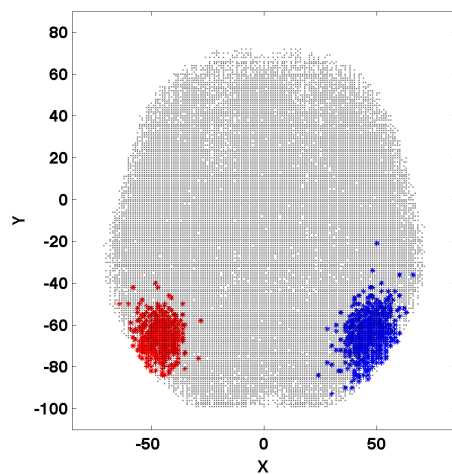
Topic 51
motor
finger
movements
stimulation
bimanual
sensorimotor
movement
alpha
handed
dominant



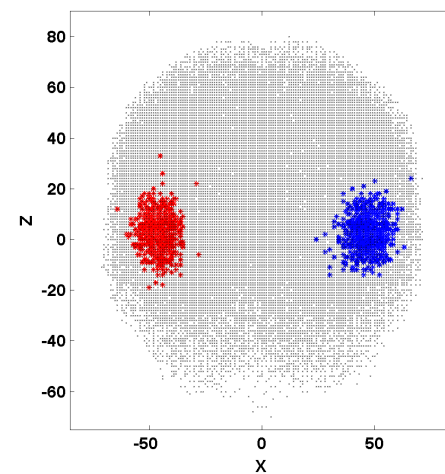
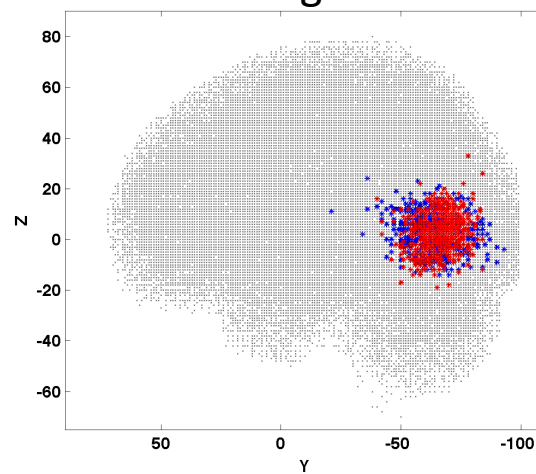
Rank 30

Topic 50

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

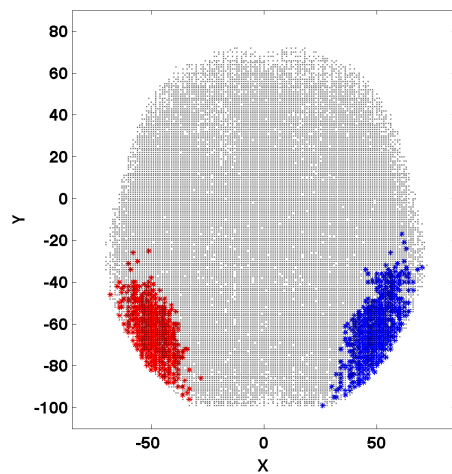


Training Set 1

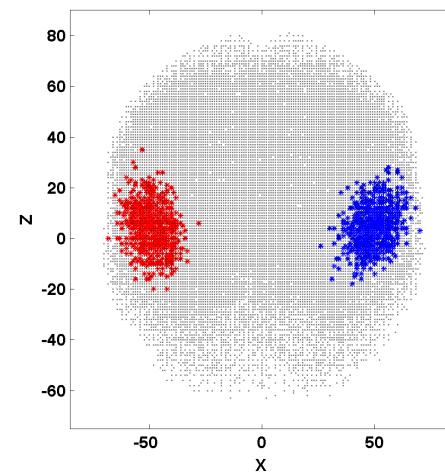
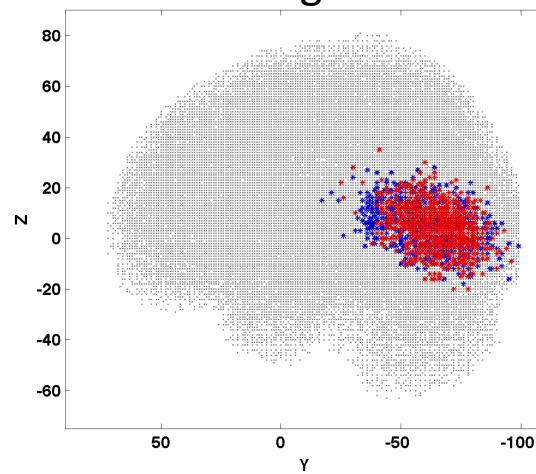


Topic 88

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

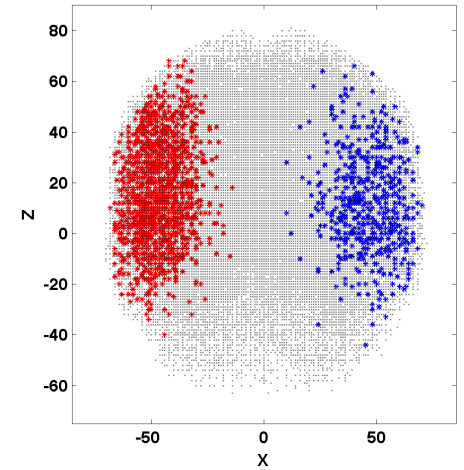
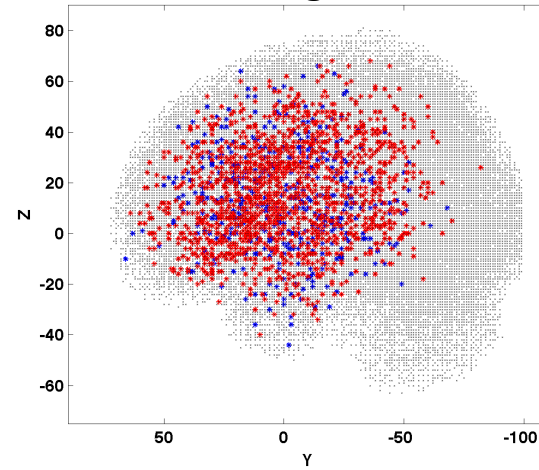
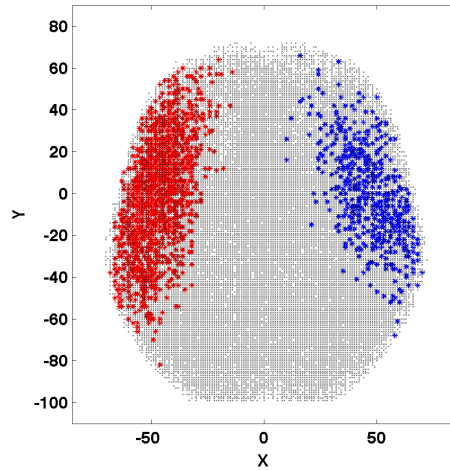


Training Set 2



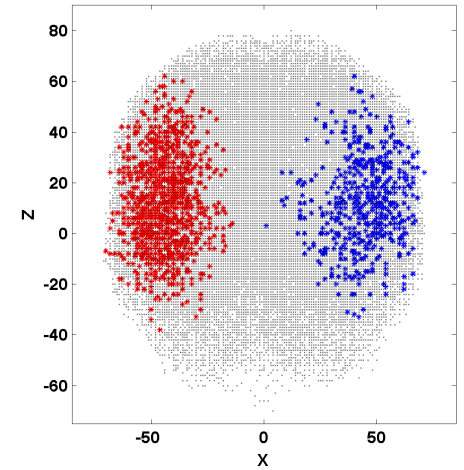
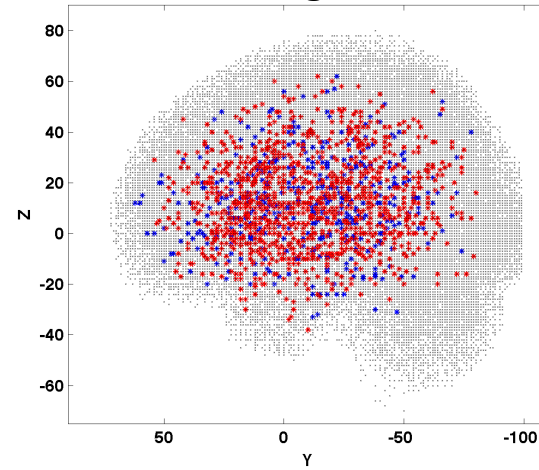
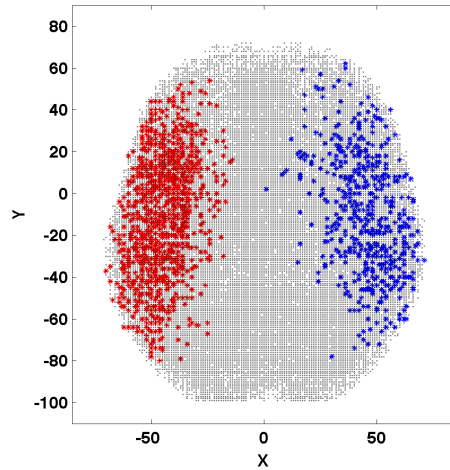
Rank 50

Training Set 1



Topic 65
words
semantic
word
lexical
meaning
semantic_processing
picture
competition
language
semantically

Training Set 2



Topic 84
language
native
chinese
languages
hearing
bilinguals
linguistic
japanese
deaf
speakers

Stability analysis

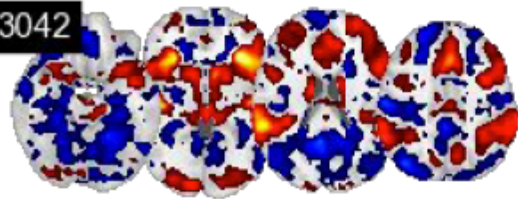
- Approximately 50% of topics are stable, independent of the training data

Image Decoding

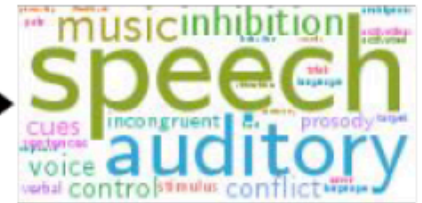
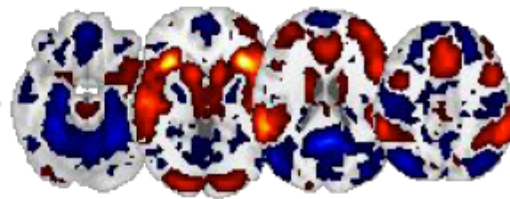
Original map

Reconstructed map

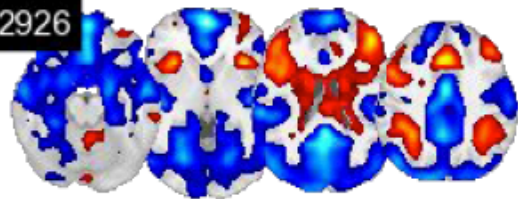
3042



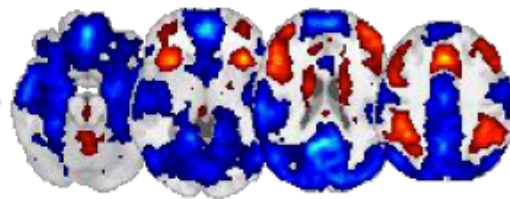
$R^2 = 0.70$



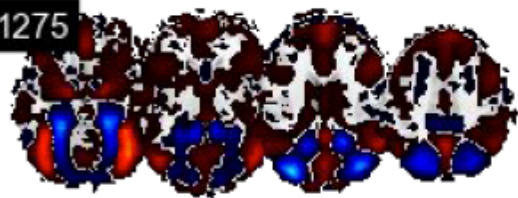
2926



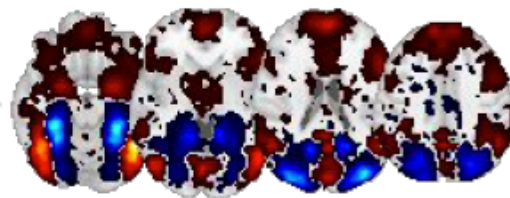
$R^2 = 0.70$



1275



$R^2 = 0.73$



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