

# A common high-dimensional linear model of representational spaces in human cortex

Jim Haxby

Center for Cognitive Neuroscience, Dartmouth College  
Center for Mind/Brain Sciences (CIMeC), University of Trento



Supported by NSF CRCNS German-US Collaboration

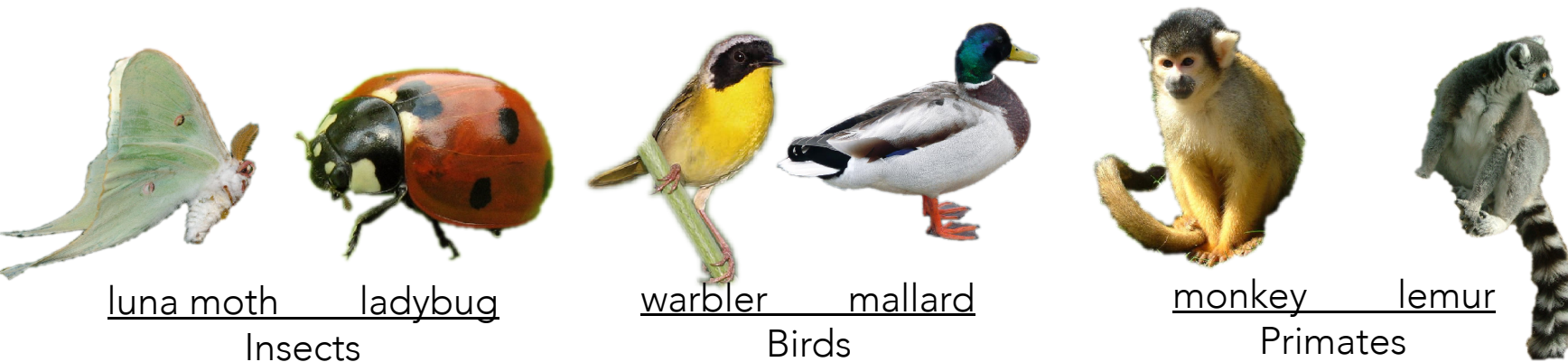
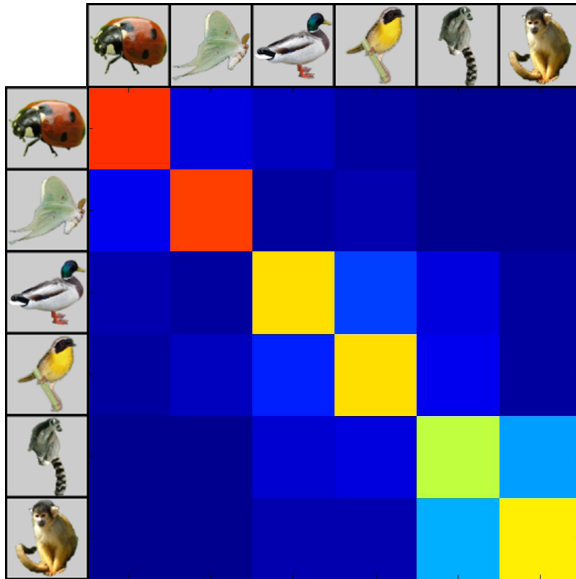
# Modeling representational spaces in human cortex

- MVPA – decoding population responses from fMRI
- Hyperalignment – building a model bases on tuning functions that are shared across brains
- HyperCortex – proposal for a functional atlas based on a common, high-dimensional model of representational spaces in human cortex

## MVPA:

# Decoding fine-grained distinctions from fine-scale patterns

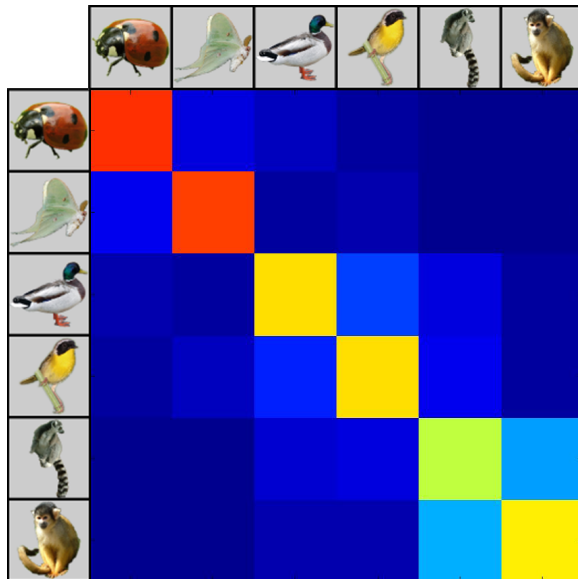
Within-subject classification  
(new model for each subject)



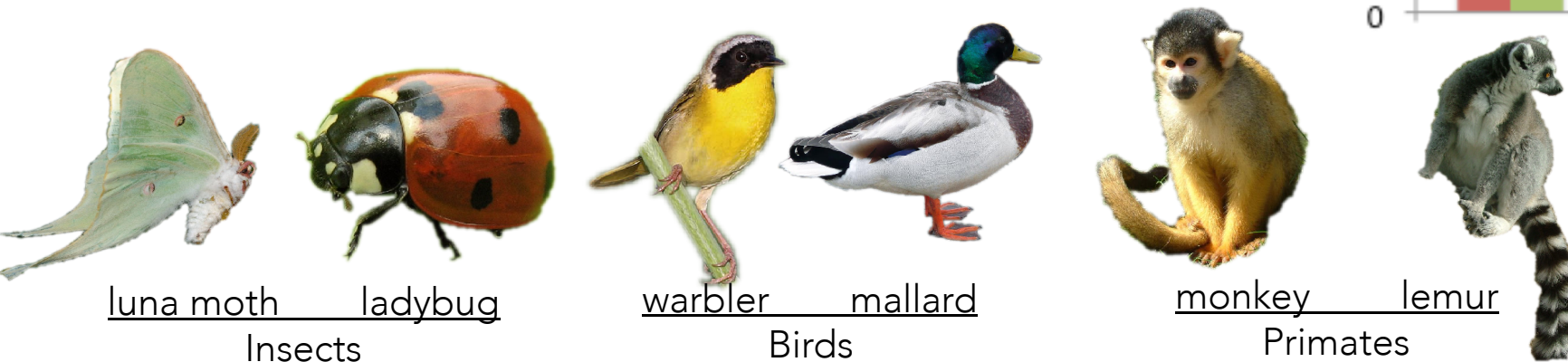
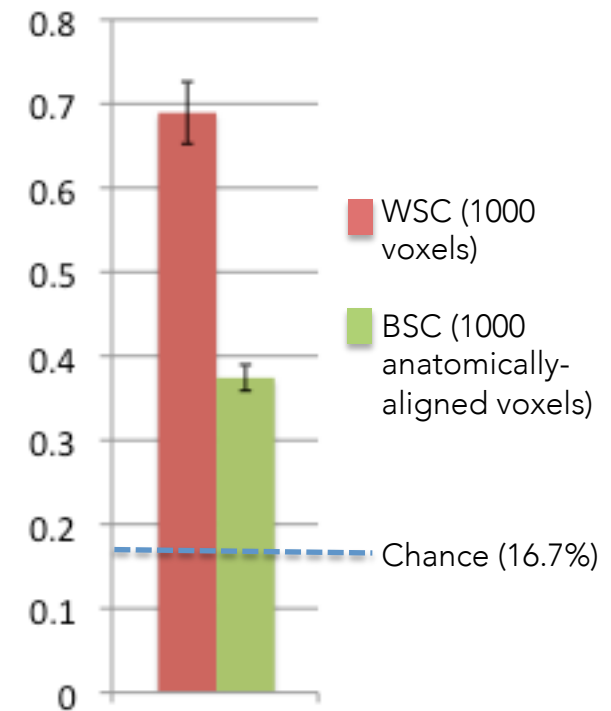
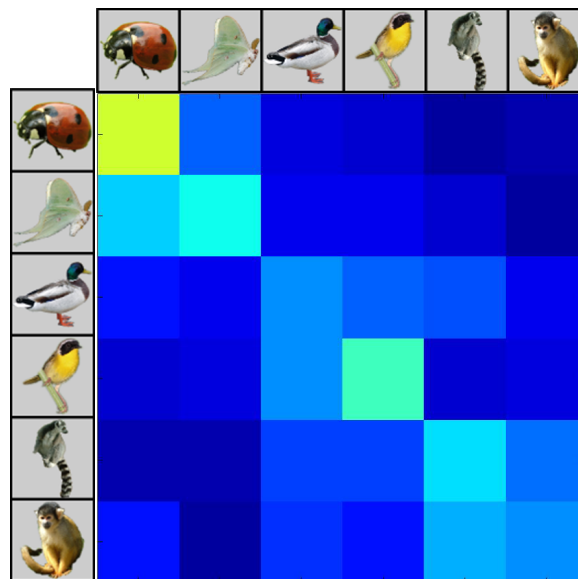
(Haxby et al. 2011; Connolly et al. 2012)

# MVPA – The problem: Fine-scale patterns are individual-specific

Within-subject classification  
(new model for each subject)



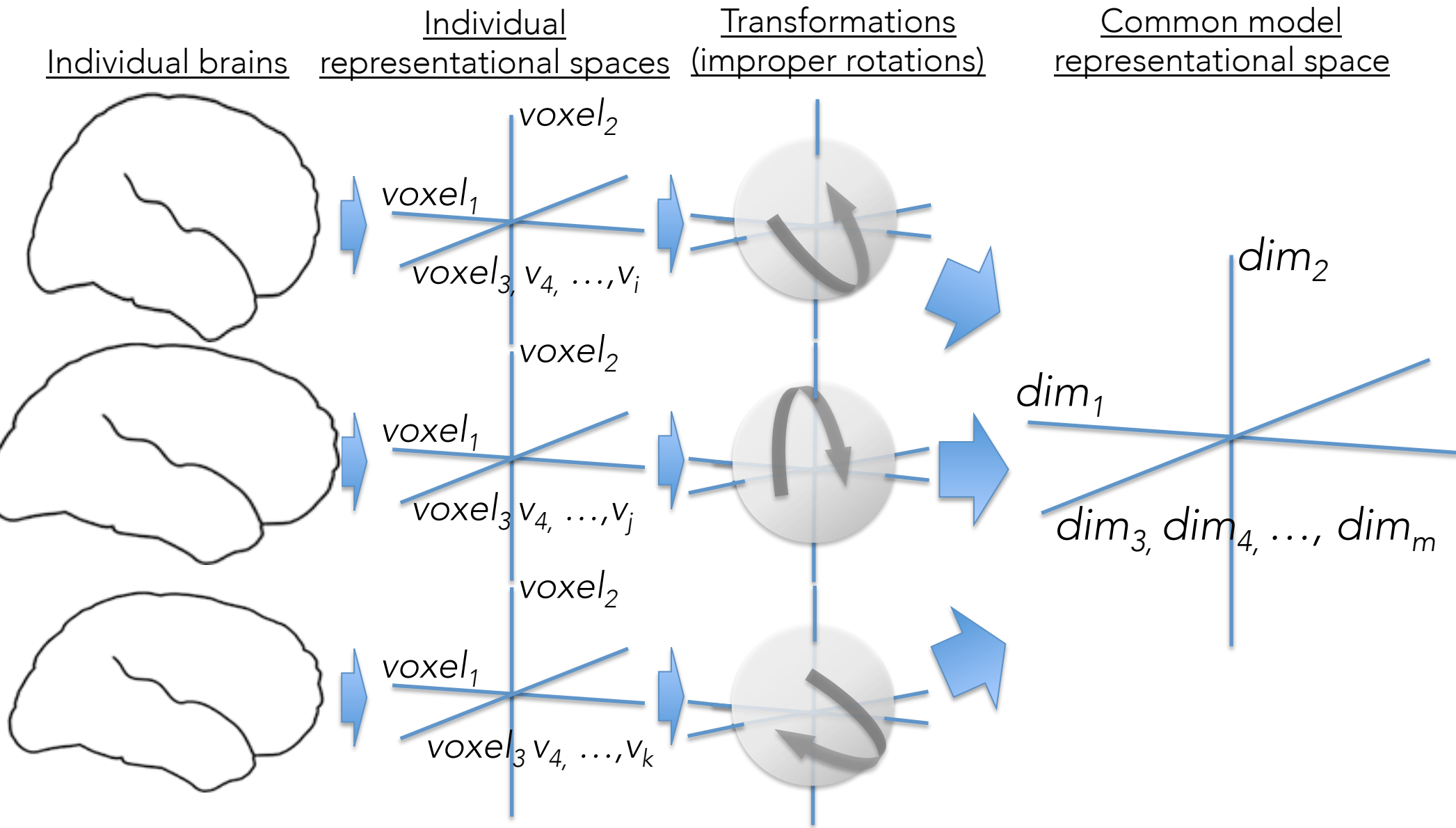
Between-subject classification  
(common model based on anatomy)



(Haxby et al. 2011; Connolly et al. 2012)

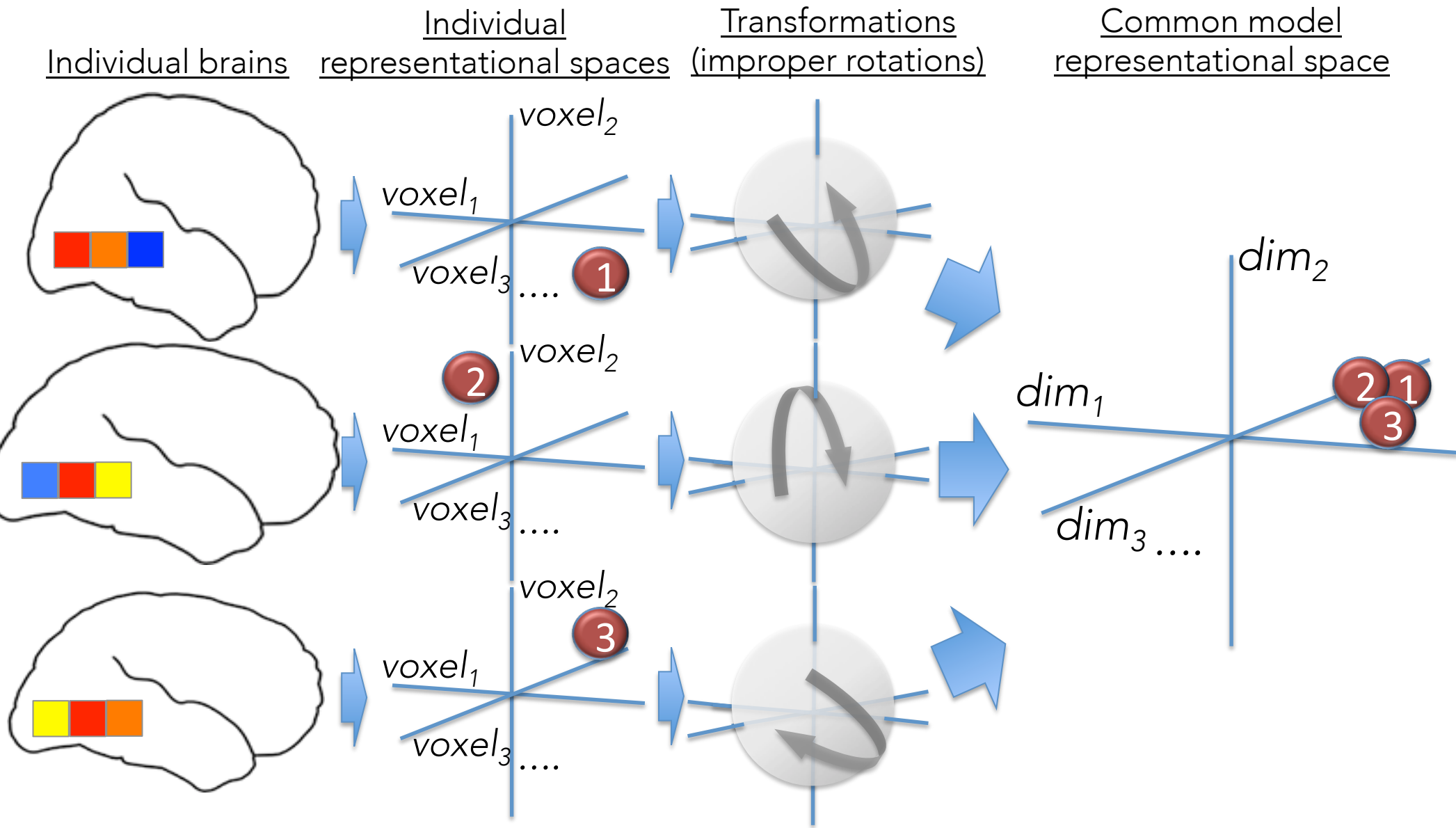
# Hyperalignment:

Individual representational spaces  $\Leftrightarrow$  common representational space



# Hyperalignment:

Individual representational spaces  $\rightleftharpoons$  common representational space



Hyperalignment parameters are estimated from responses recorded during movie viewing



*Raiders of the Lost Ark*

*Life on Earth*



*The Wire*

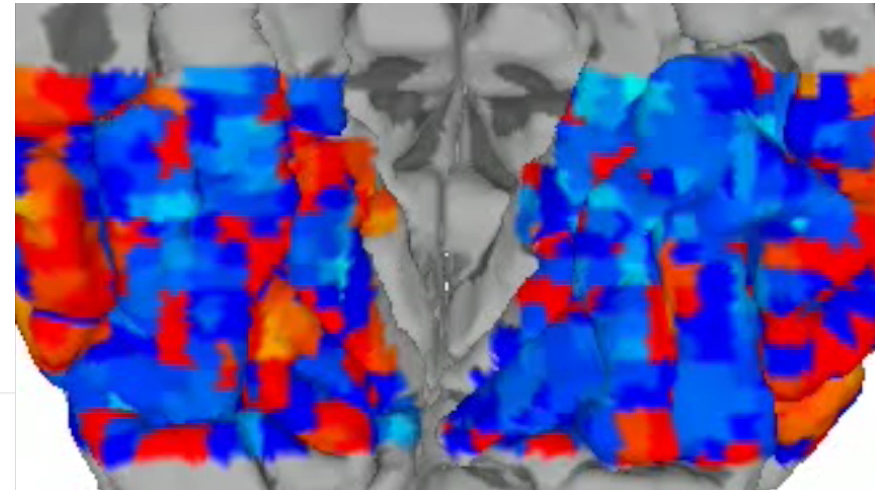
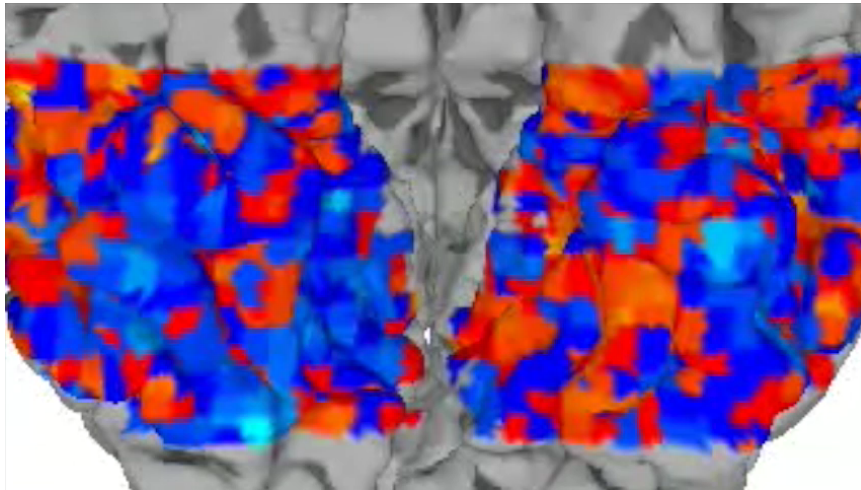


# Broad sampling of a neural representational space with a movie

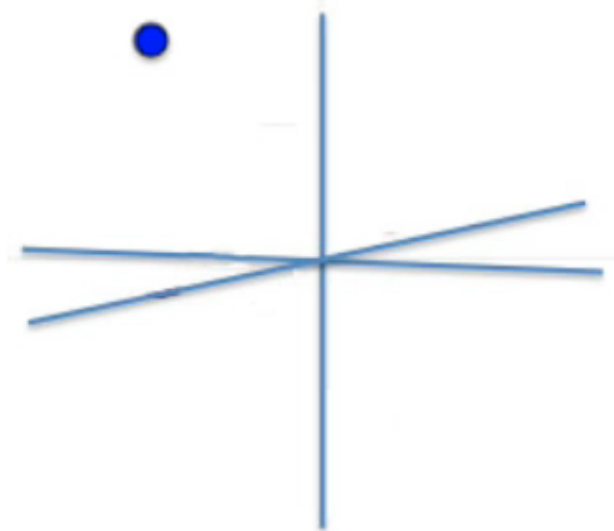
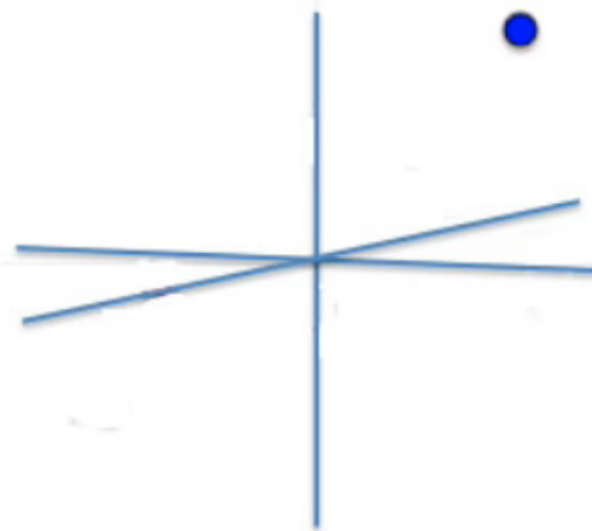
S1

S2

Response  
patterns in  
cortex

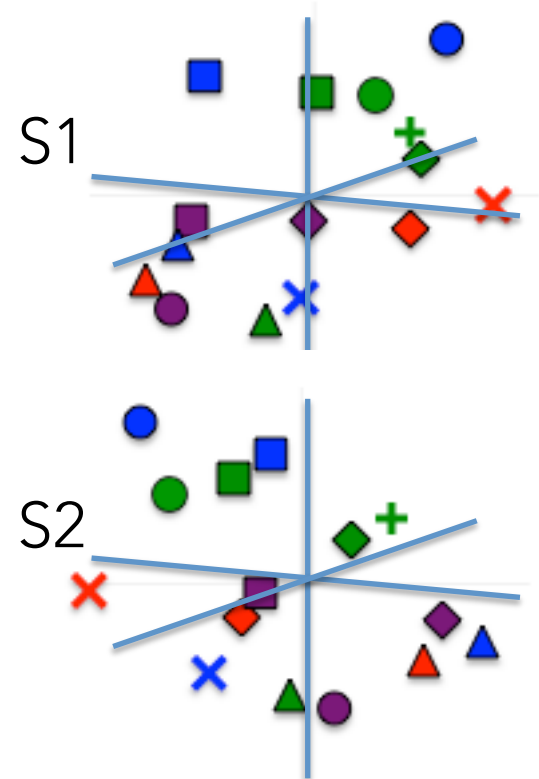


15 response  
pattern vectors  
in individual 3D  
representational  
spaces  
(full exp't has  
>2600 vectors in  
>50,000D space)

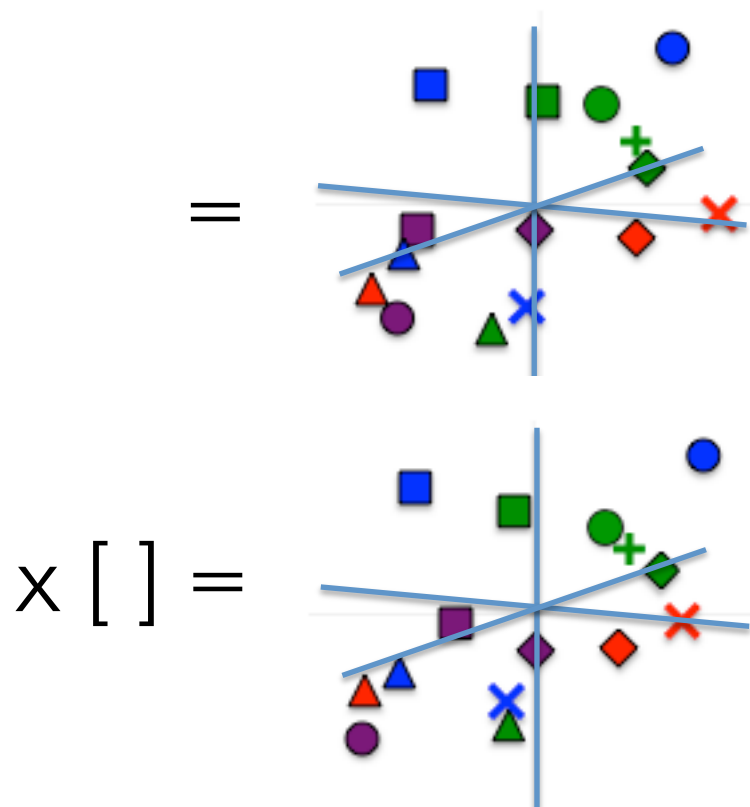




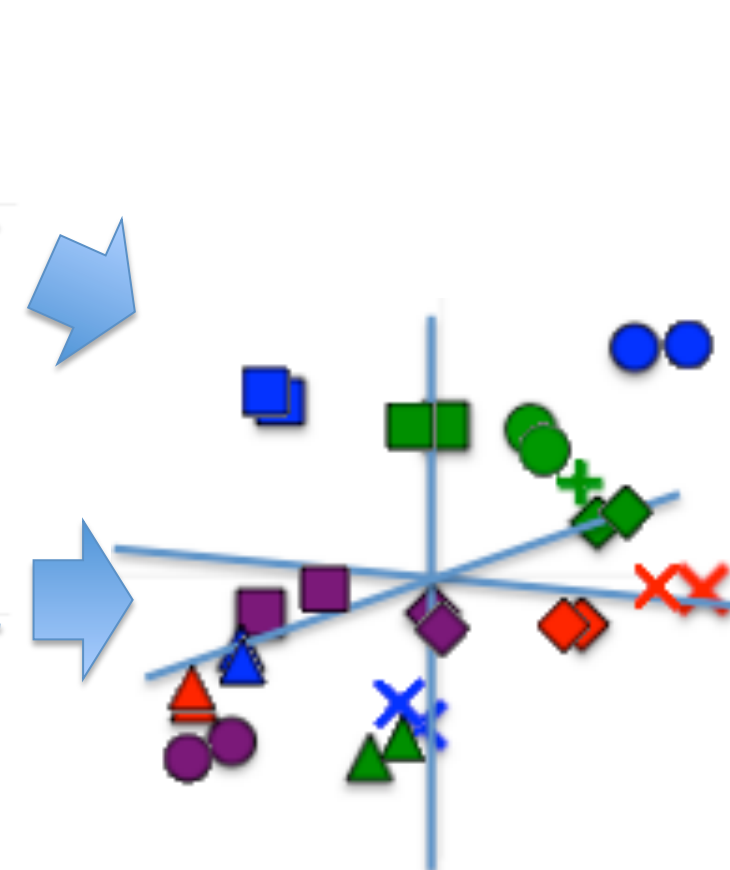
Individual  
representational spaces



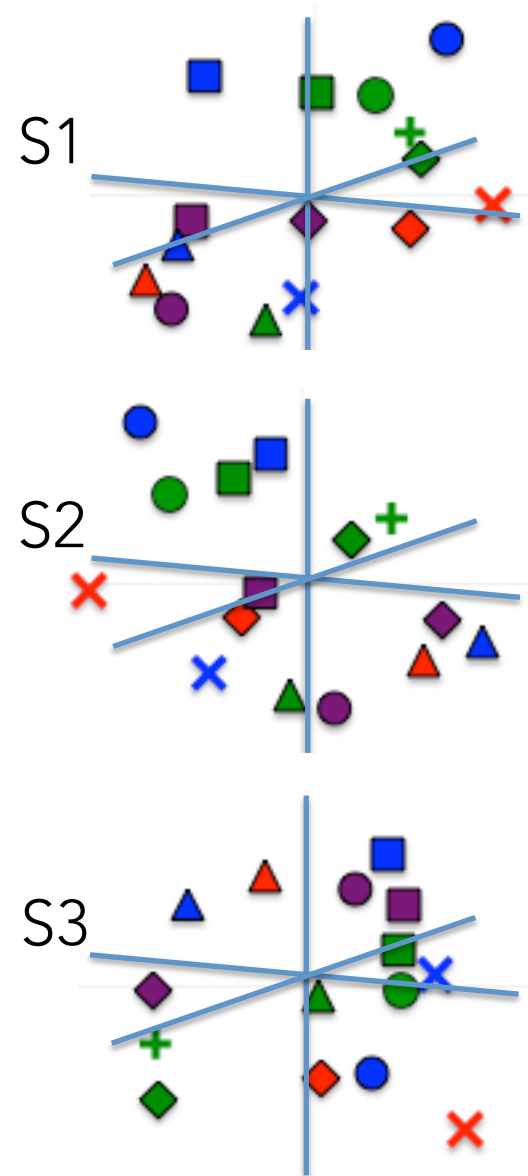
Procrustes transformations  
(improper rotations)



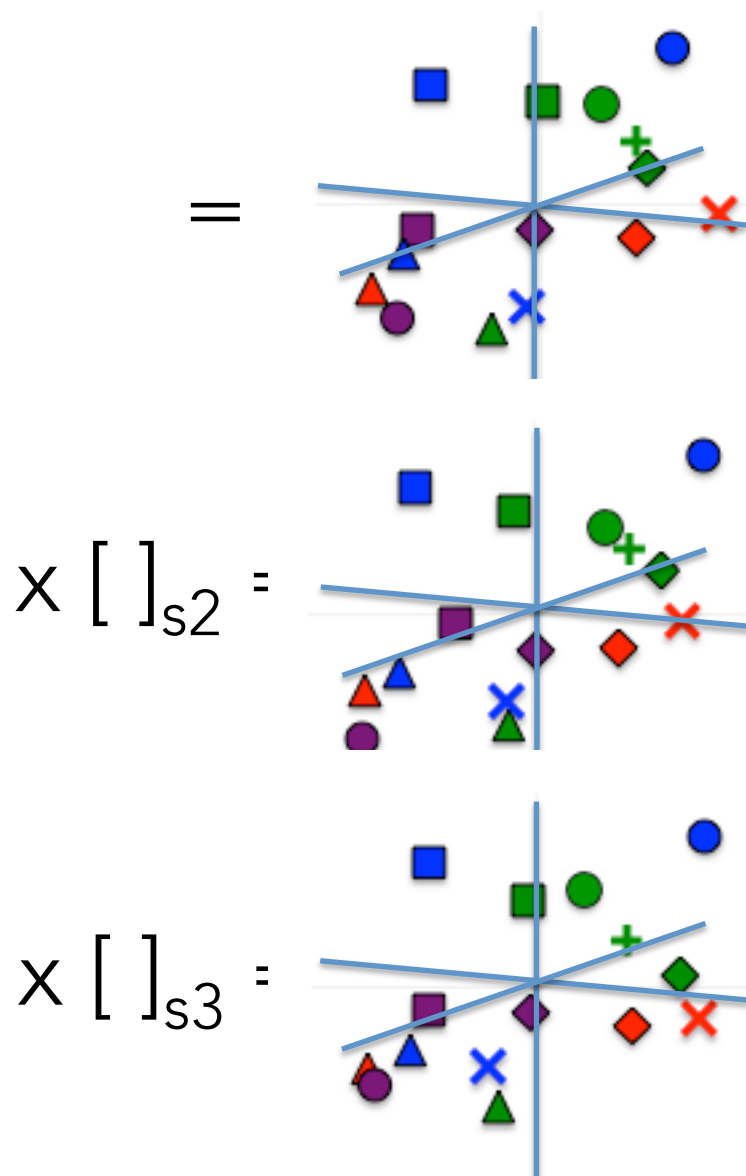
Common model  
representational space



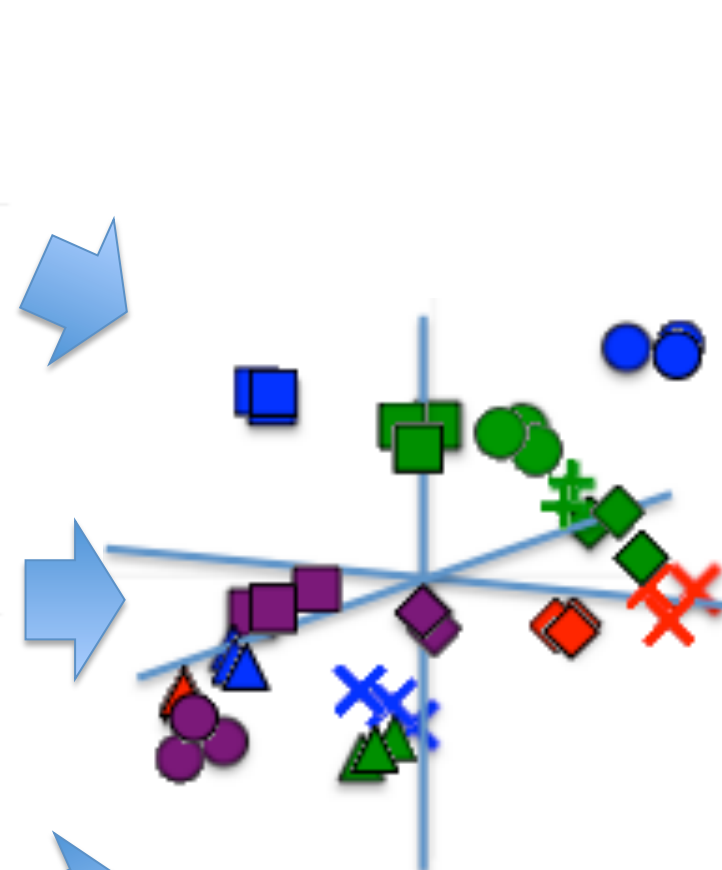
Individual  
representational spaces



Procrustes transformations  
(improper rotations)

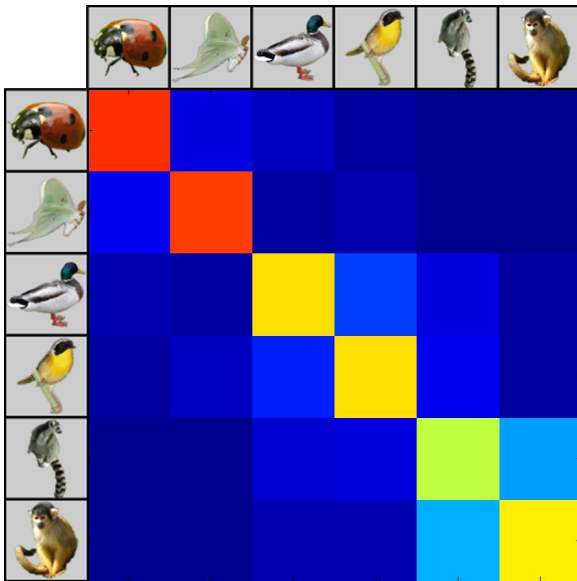


Common model  
representational space

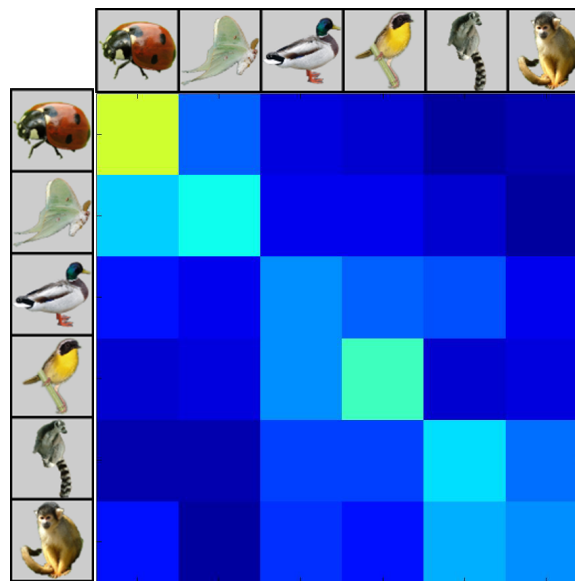


# MVPA – The problem: Fine-scale patterns are individual-specific

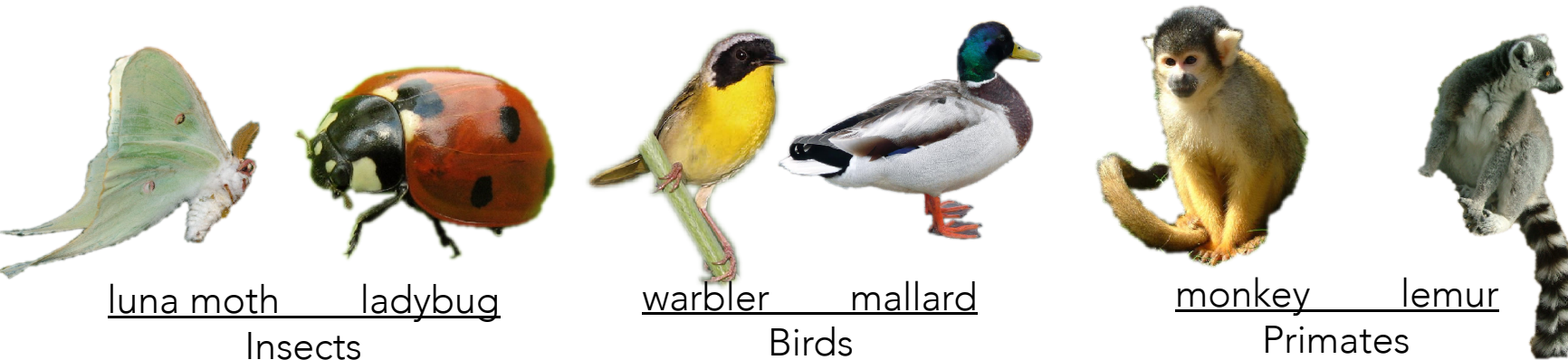
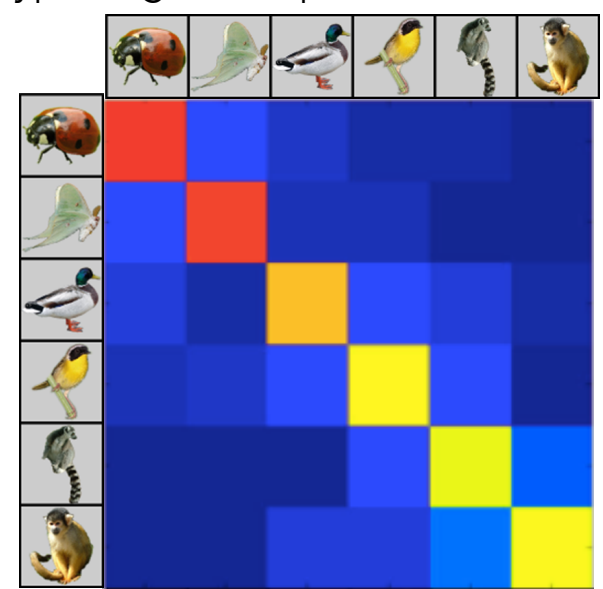
Within-subject classification  
new model for each subject



Between-subject classification  
common model based on anatomy



common model using movie-based  
hyperalignment parameters

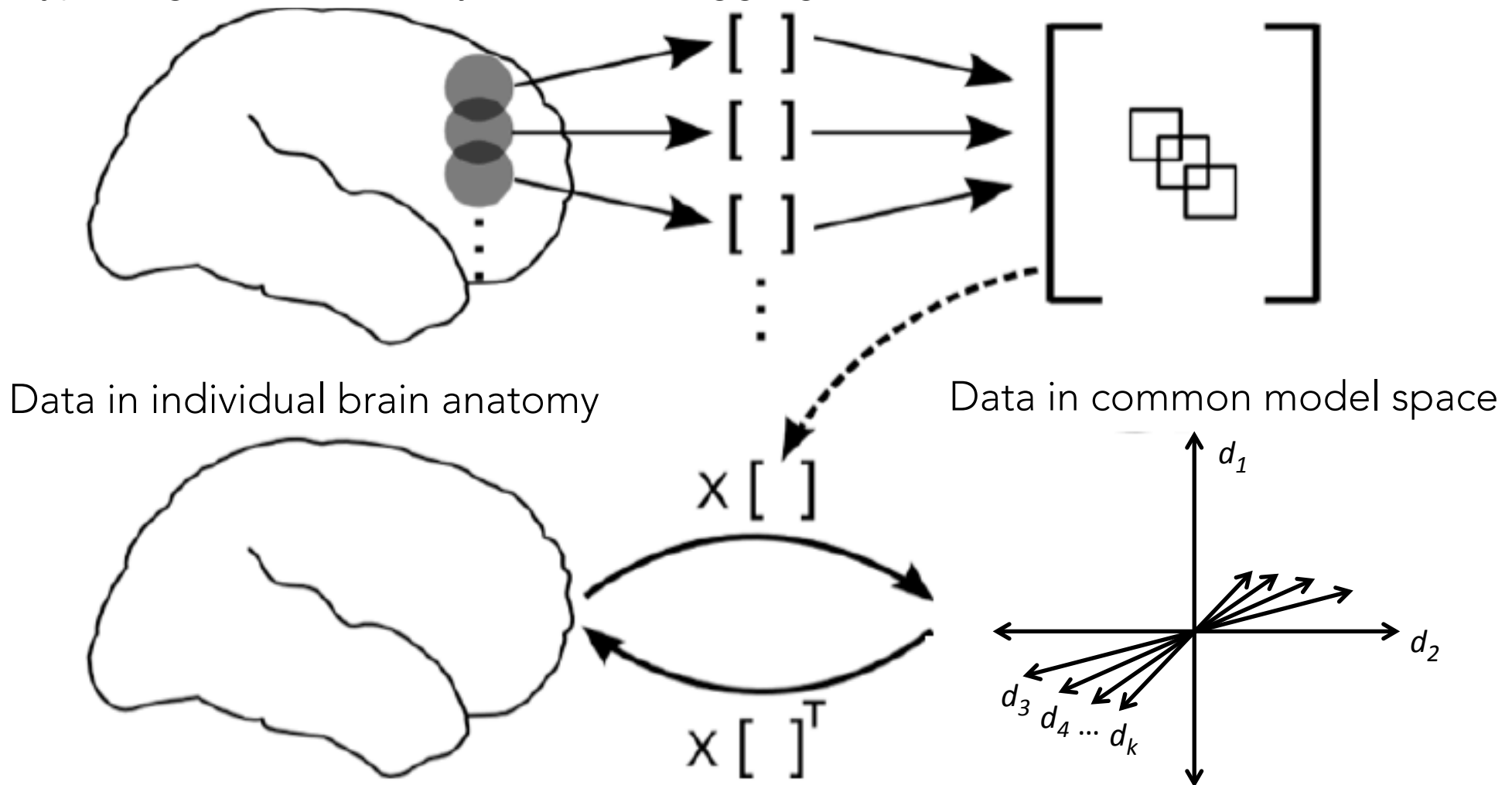


(Haxby et al. 2011; Connolly et al. 2012)

# Modeling representational spaces in all human cortex with searchlight hyperalignment

Voxels in overlapping searchlights are hyperaligned across subjects

Overlapping searchlight transformation matrices are aggregated into a whole cortex matrix

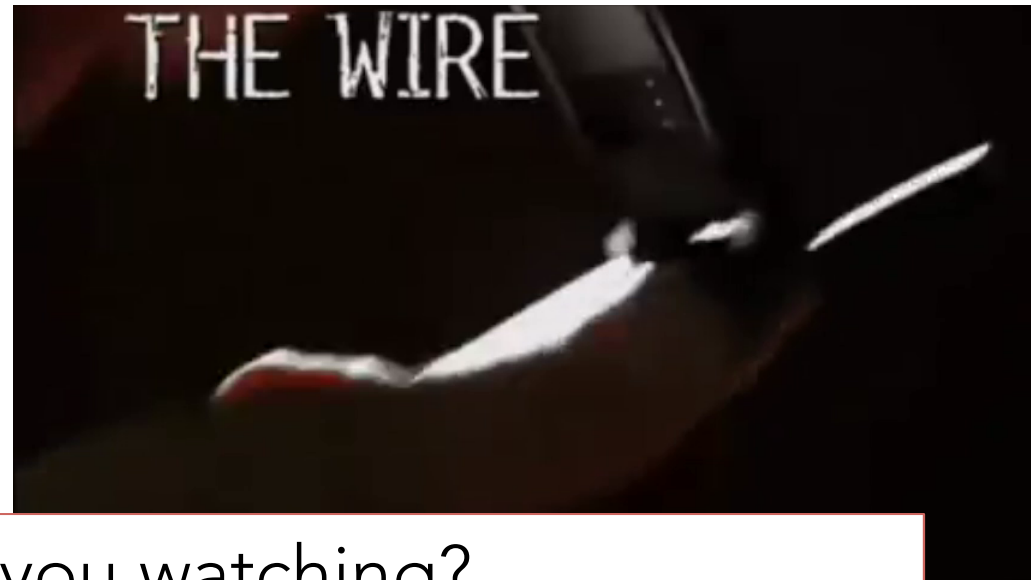


Hyperalignment parameters are estimated from responses recorded during movie viewing



*Raiders of the Lost Ark*

*Life on Earth*

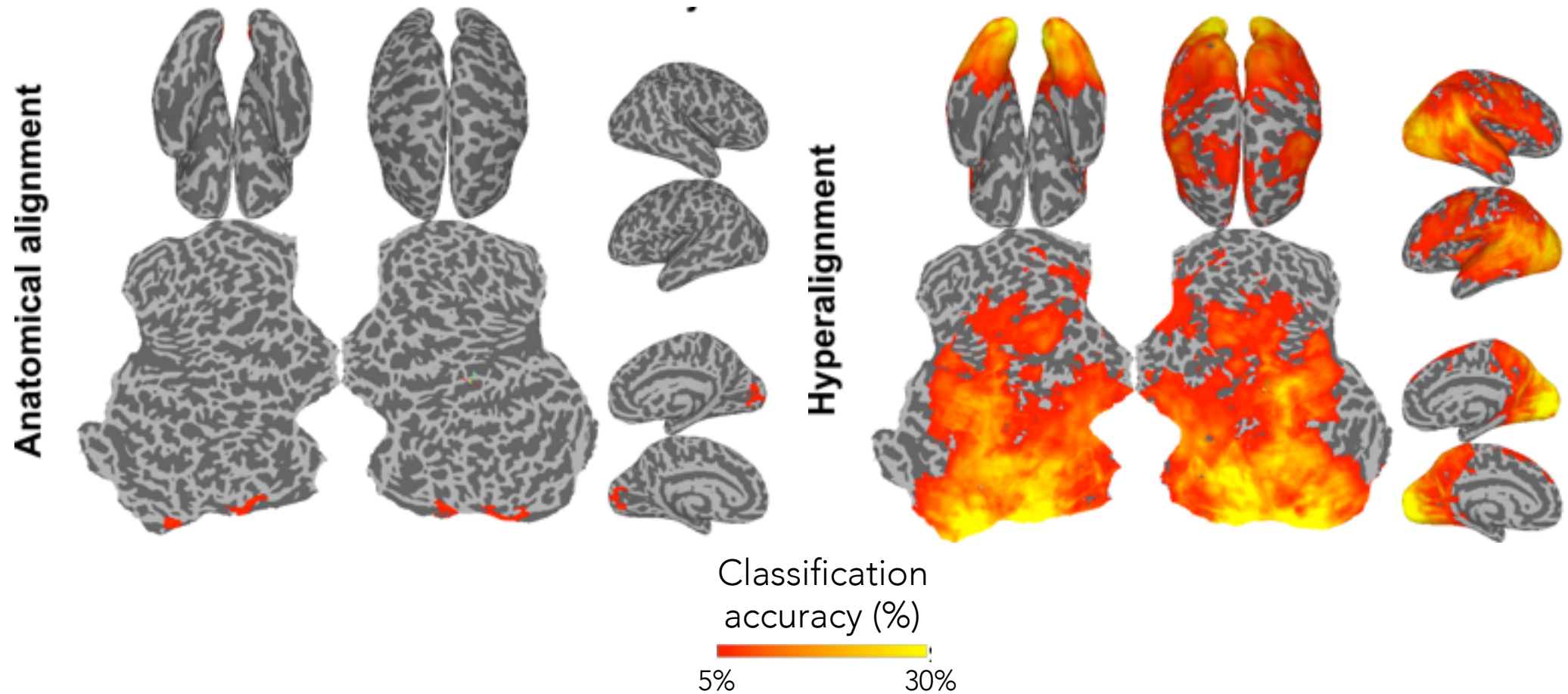


*ire*

What part of the movie are you watching?

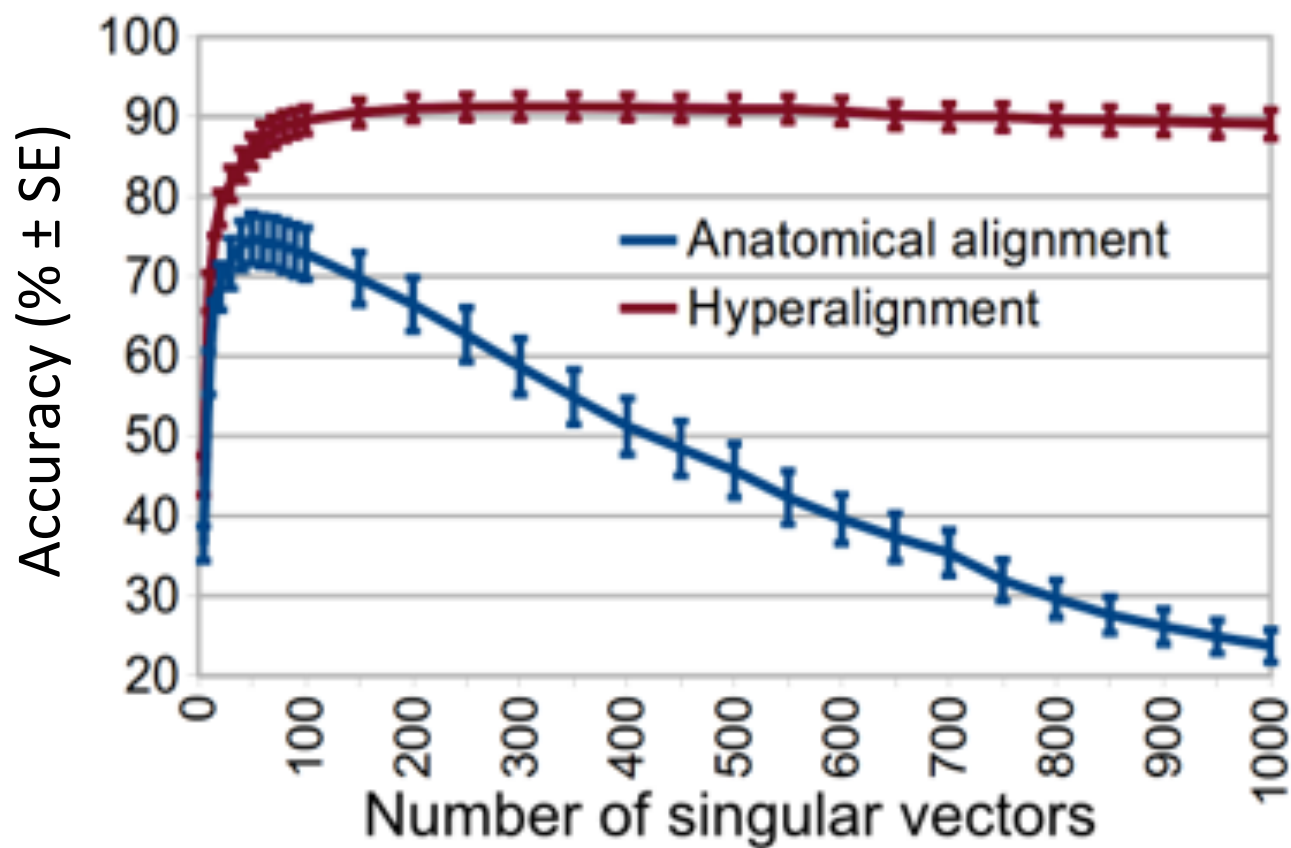
From brain activity (fMRI), we can decode which 15 sec segment you are watching with  $>90\%$  accuracy

Whole-brain hyperalignment affords between-subject classification of 15 s movie time segments in occipital, temporal, parietal, and frontal cortices

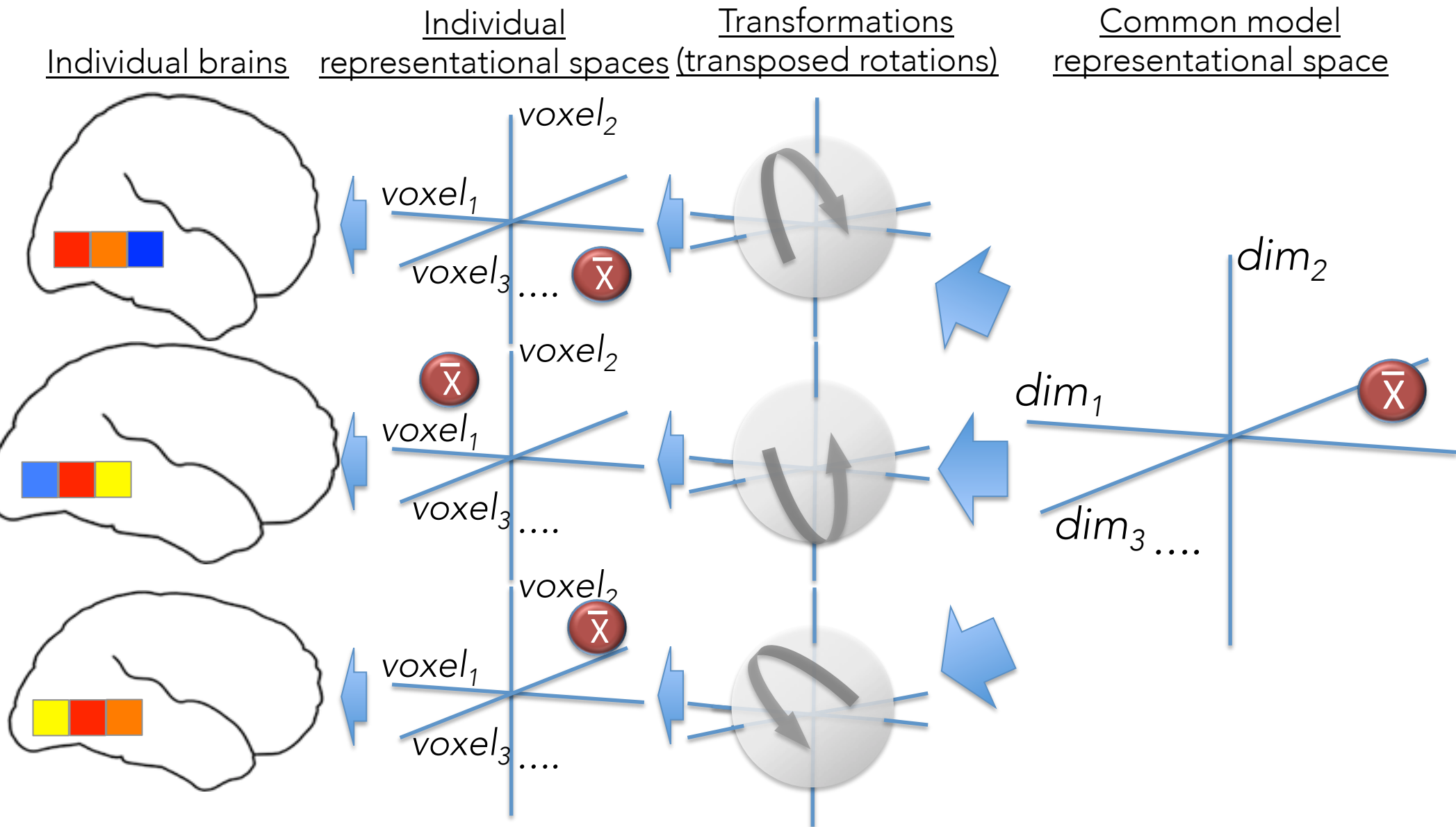




Whole-brain hyperalignment increases between-subject classification of 15 s movie time segments for the whole brain (after SVD dimensionality reduction)

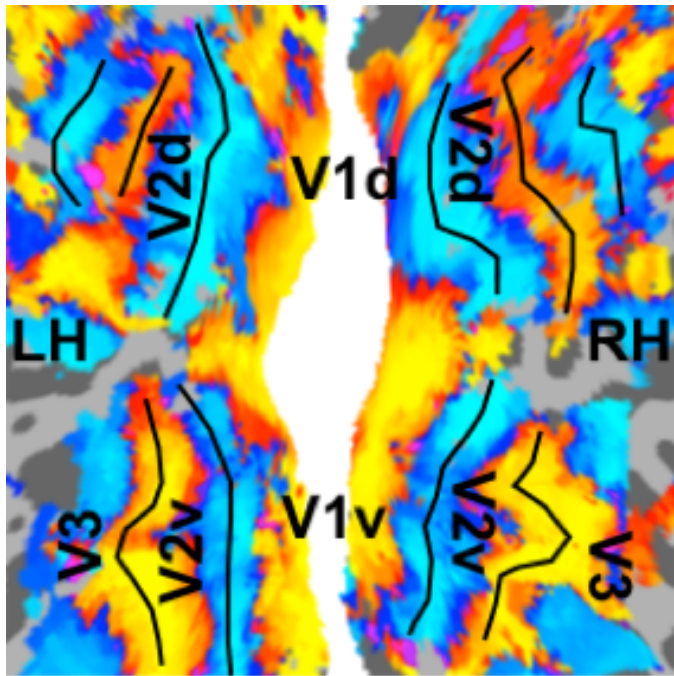


# Projecting group data from common model space into individual subject's anatomy

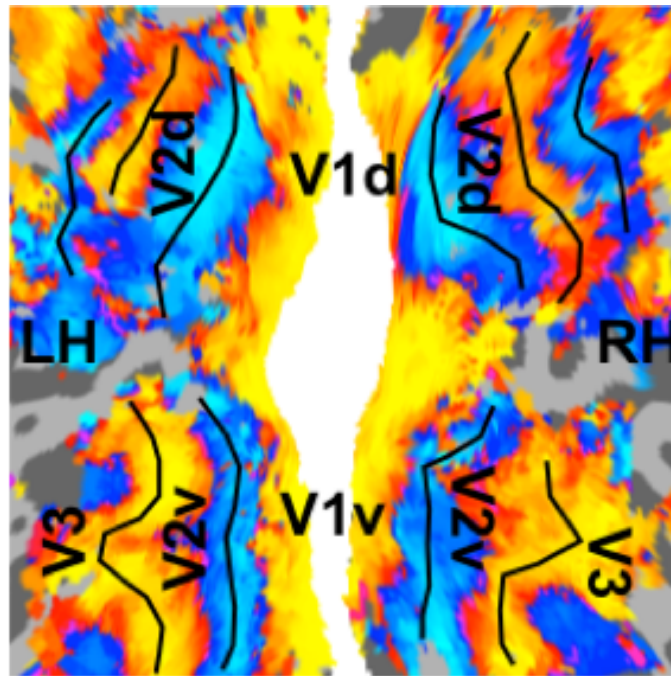


# Mapping retinotopy by projecting other subjects' polar angle maps into a different subject's occipital topography

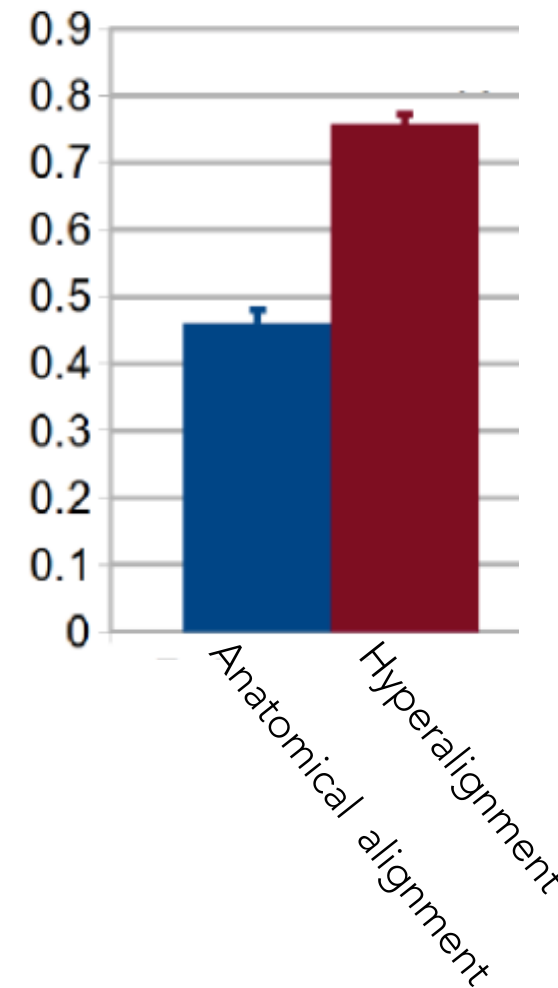
Polar angle from subject's own retinotopy data



Polar angle from other subjects' retinotopy data



Correlation between measured and projected



Can a high-dimensional common model of human cortex be leveraged to build a new type of functional brain atlas?

Brain atlases are an essential tool for functional neuroimaging research

- Provide a common basis for reporting results
- Allow comparisons across studies affording
  - Replication testing
  - Interpretation
  - Meta-analysis
- More generally, afford accrual of knowledge about the functional organization of the human brain

# Functional Brain Atlas: Current State of the Art

Results are reported in tables with anatomical x,y,z coordinates

**Table 1. Group-Average Activation for the Biological Motion Display**

Region	Talairach Coordinates			Mean T	mm <sup>3</sup>
	X	Y	Z		
R. ITS	44	−69	−7	4.63	2101
R. ITS	52	−51	3	4.49	920
R. STS	57	−41	21	4.17	112
R. Fusiform	36	−39	−19	4.72	748
R. Fusiform	34	−68	−18	4.32	174
R. Post. Occipital	14	−96	−4	4.43	151
L. ITS	−42	−70	−4	4.59	2544
L. Supramarginal	−56	−39	25	4.37	388

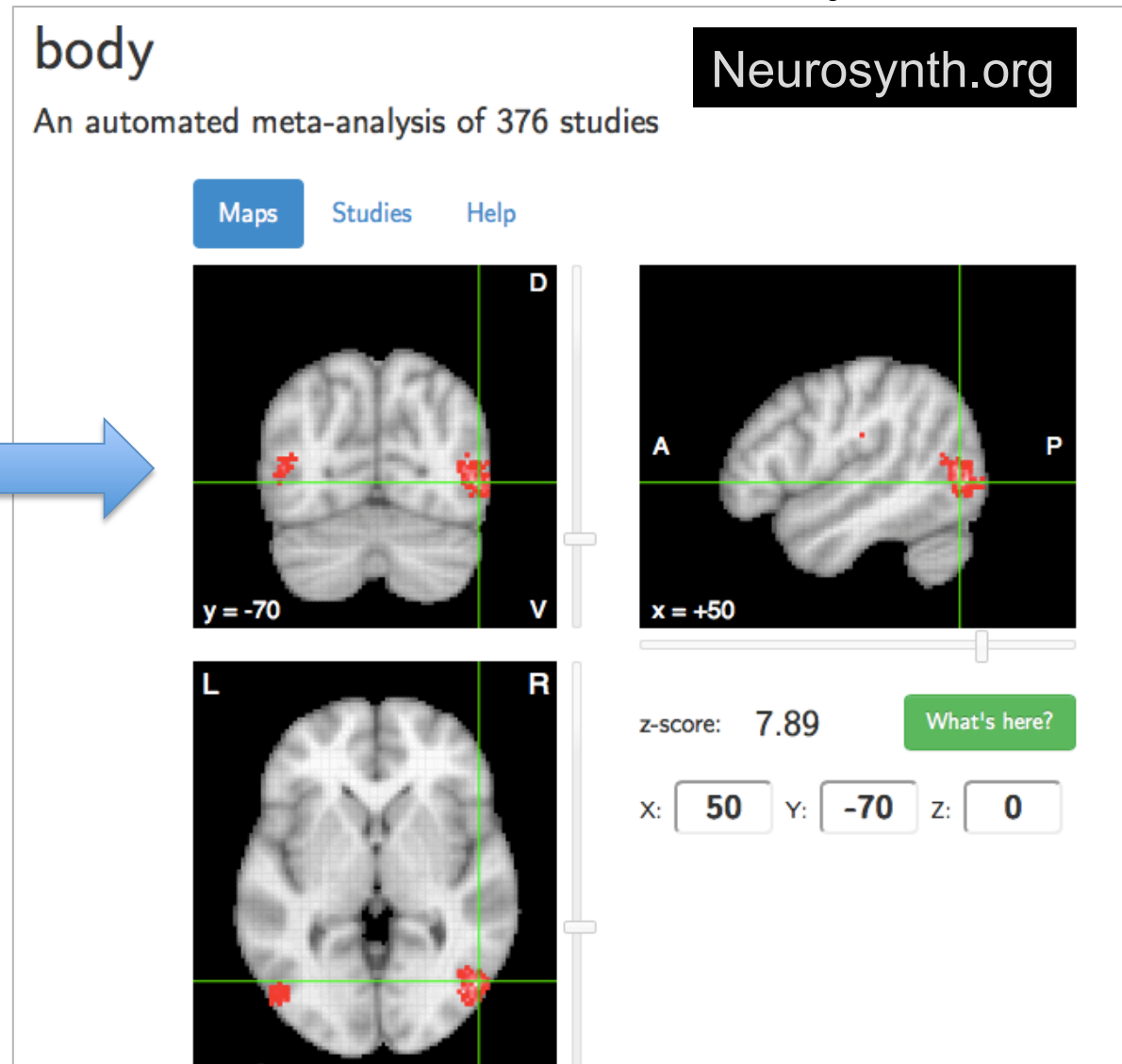
from Peelen & Downing, Neuron, 2006

# Functional Brain Atlas: Current State of the Art

Results are aggregated across studies based on x,y,z coordinates

Table 1. Group-Average Activation for the Biological Motion Display

Region	Talairach Coordinates			Mean T	mm <sup>3</sup>
	X	Y	Z		
R. ITS	44	-69	-7	4.63	2101
R. ITS	52	-51	3	4.49	920
R. STS	57	-41	21	4.17	112
R. Fusiform	36	-39	-19	4.72	748
R. Fusiform	34	-68	-18	4.32	174
R. Post. Occipital	14	-96	-4	4.43	151
L. ITS	-42	-70	-4	4.59	2544
L. Supramarginal	-56	-39	25	4.37	388





# Functional Brain Atlas: Current State of the Art

## The function of a locus is described as a “word-cloud”

Table 1. Group-Average Activation for the Biological Motion Display

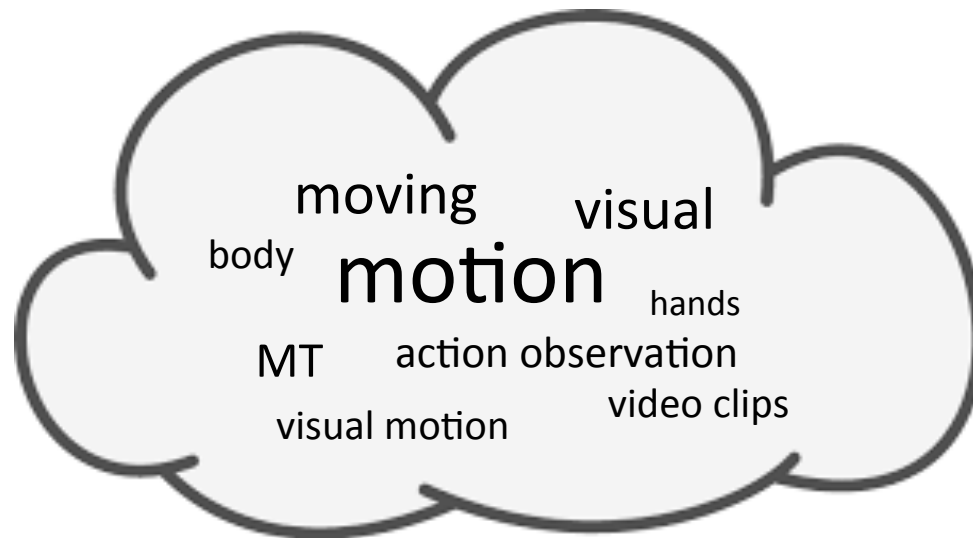
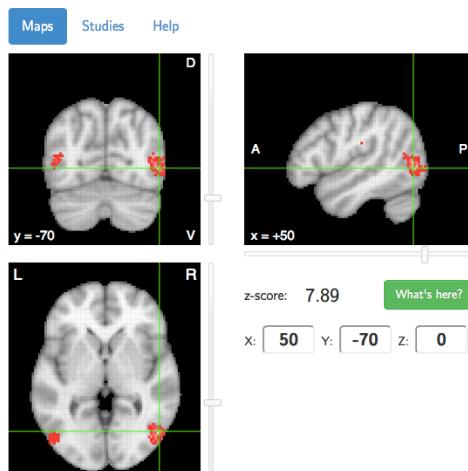
Region	Talairach Coordinates			Mean T	mm <sup>3</sup>
	X	Y	Z		
R. ITS	44	-69	-7	4.63	2101
R. ITS	52	-51	3	4.49	920
R. STS	57	-41	21	4.17	112
R. Fusiform	36	-39	-19	4.72	748
R. Fusiform	34	-68	-18	4.32	174
R. Post. Occipital	14	-96	-4	4.43	151
L. ITS	-42	-70	-4	4.59	2544
L. Supramarginal	-56	-39	25	4.37	388



body

An automated meta-analysis of 376 studies

Neurosynth.org



# Functional Brain Atlas: Current State of the Art

## The function of a locus is described as a “word-cloud”

Table 1. Group-Average Activation for the Biological Motion Display

Region	Talairach Coordinates			Mean T	mm <sup>3</sup>
	X	Y	Z		
R. ITS	44	-69	-7	4.63	2101
R. ITS	52	-51	3	4.49	920
R. STS	57	-41	21	4.17	112
R. Fusiform	38	-39	-19	4.72	748
R. Fusiform	34	-68	-18	4.32	174
R. Post. Occipital	-1	-96	-4	4.43	351
L. ITS	44	-69	-7	4.63	2101
L. Supramarginal	-56	-39	25	4.37	388

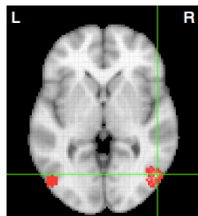
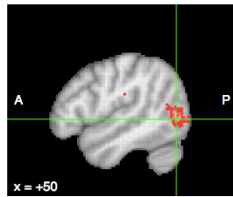
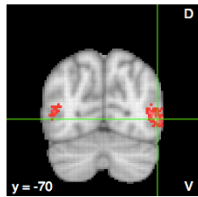
Why are anatomical coordinates inadequate for capturing neural representation?

body

An automated meta-analysis of 376 studies

Neurosynth.org

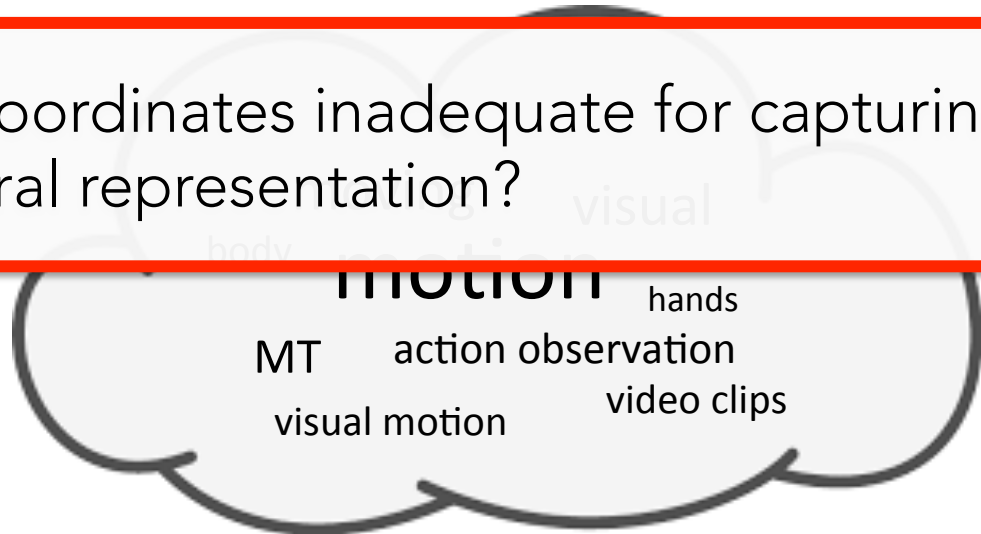
Maps Studies Help



z-score: 7.89

What's here?

X: 50 Y: -70 Z: 0



## Why are anatomical coordinates inadequate for capturing neural representation?

- Response tuning functions for voxels with the same anatomical coordinates are highly variable across brains.
- The basic unit for neural representation is the population response, not the responses of single voxels (or single neurons).

# HyperCortex

## Proposal for a new functional brain atlas

based on a high-dimensional common representational space

- Model dimensions have response tuning functions that are highly similar across brains.
- Brain responses are captured as pattern vectors, reflecting population codes with response basis functions that are shared across brains.
- Fine-scale topographies are preserved and can be recreated in each individual brain.
- Data can be shared, interpreted, and subjected to meta-analysis in a computational structure that captures fine-scale patterns of activity that encode fine distinctions.

## Some acknowledgements



Swaroop Guntupalli  
now at Caltech  
Hyperalignment development



Peter Ramadge  
Electrical Engineering  
Princeton University



Yaroslav Helchenko and Michael Hanke  
CCN at Dartmouth and the University of Magdeburg, Germany  
Software engineering



<http://neuro.debian.net>

neurodebian

