

NEURAL COMPUTATION

OR

**NEUROSCIENCE FOR
COMPUTER SCIENTISTS**

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8/31/2017

TODAY

- * Overview of class
- * What I do
- * Syllabus

BIG QUESTION

- * Given a system that computes something,
 - * what does it compute?
 - * how does it work?
 - * what are the *internal representations* that support that computation?

WHAT DOES IT MEAN TO UNDERSTAND?

- * David Marr (1945 - 1980), *Vision* (1982)
- * **Levels of analysis:**
 - * Computational
 - * Algorithmic/Representational
 - * Implementation

COMPUTING SYSTEMS

- * Biological: brains
- * Artificial: neural networks
- * Artificial: microcircuits

TOPICS

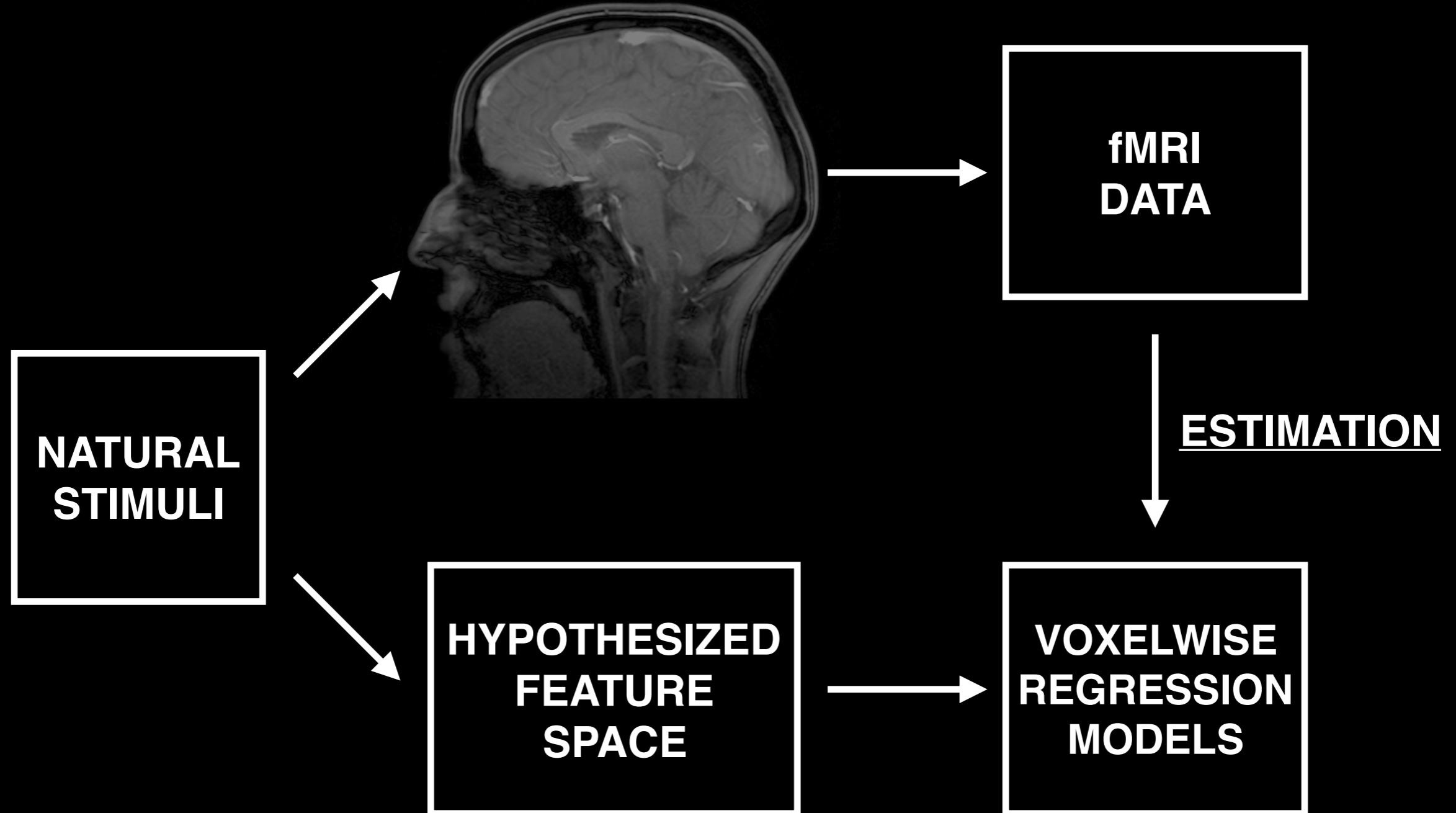
- * Brains
 - * Strategies for neuroscience / philosophies of experimental design
 - * Dealing with real, noisy data & uncertainty
 - * Fitting predictive encoding & decoding models (& examining philosophies thereof)

TOPICS

- * Artificial neural networks
 - * Strategies for gaining understanding
- * Microcircuits (ever so briefly)

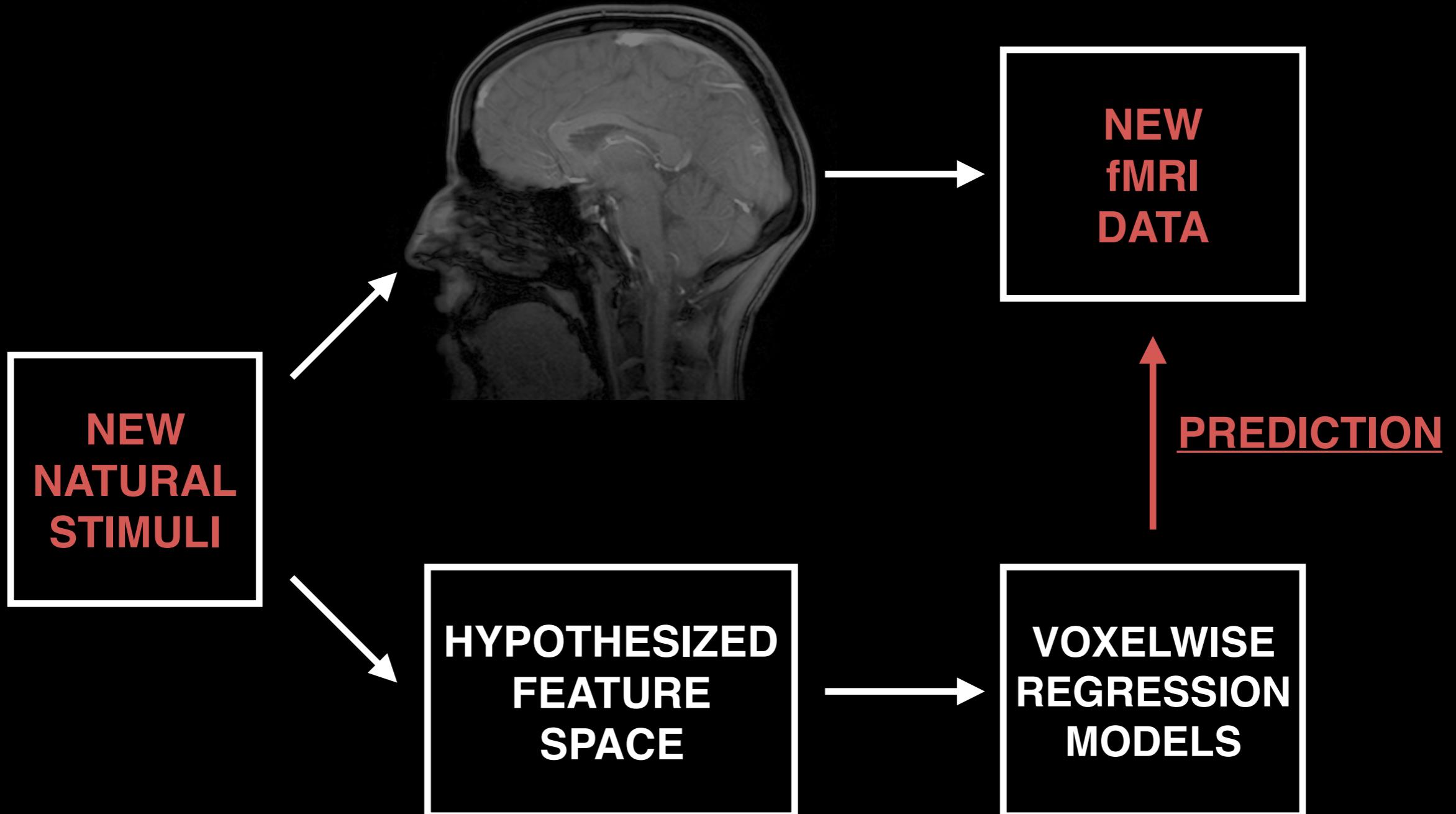
MY WORK

VOXELWISE MODELING

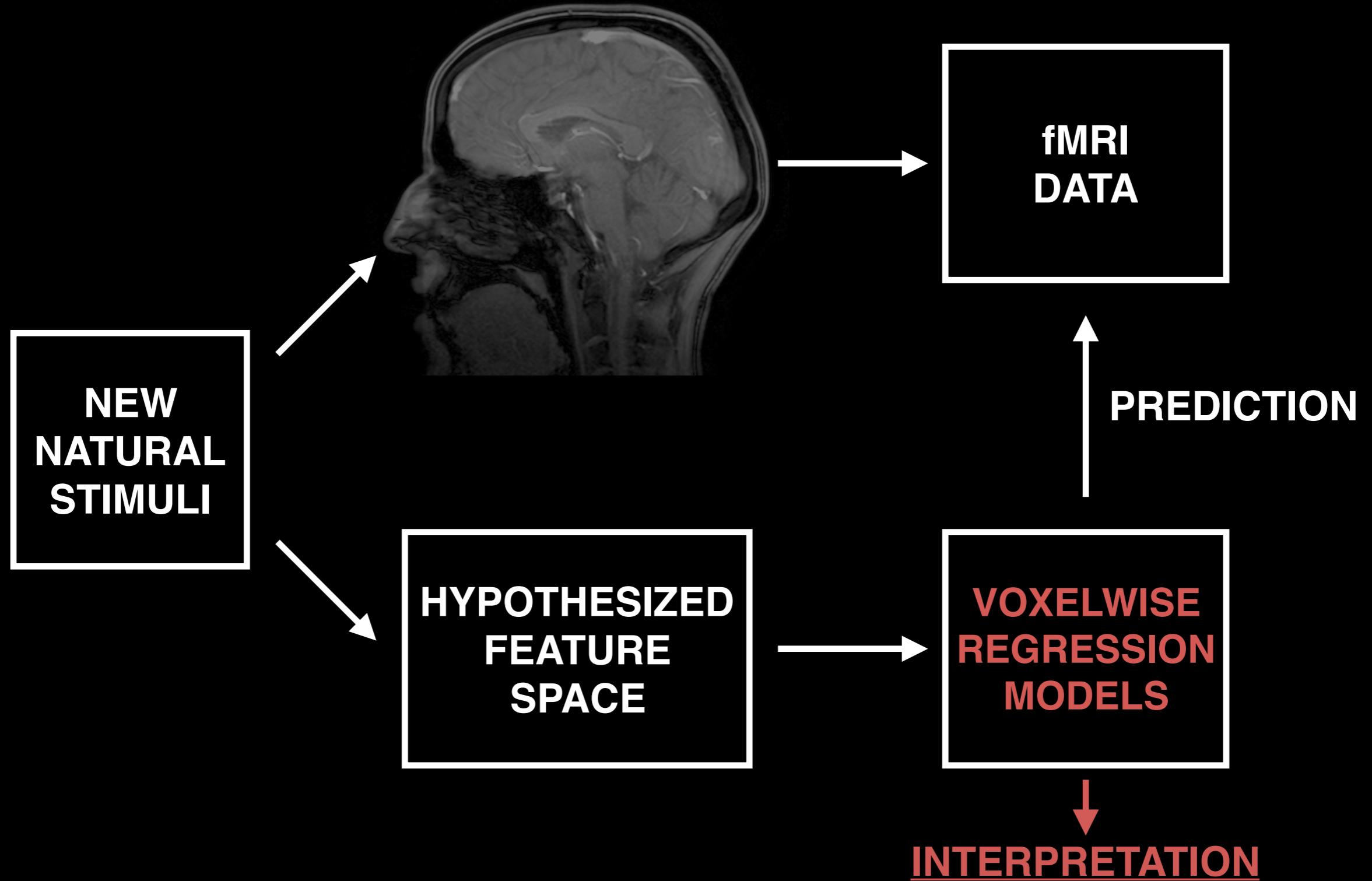


Kay et al. *Nature* (2008), Naselaris et al. *Neuron* (2009),
Nishimoto et al. *Current Biology* (2011), Huth et al. *Neuron* (2012), etc.

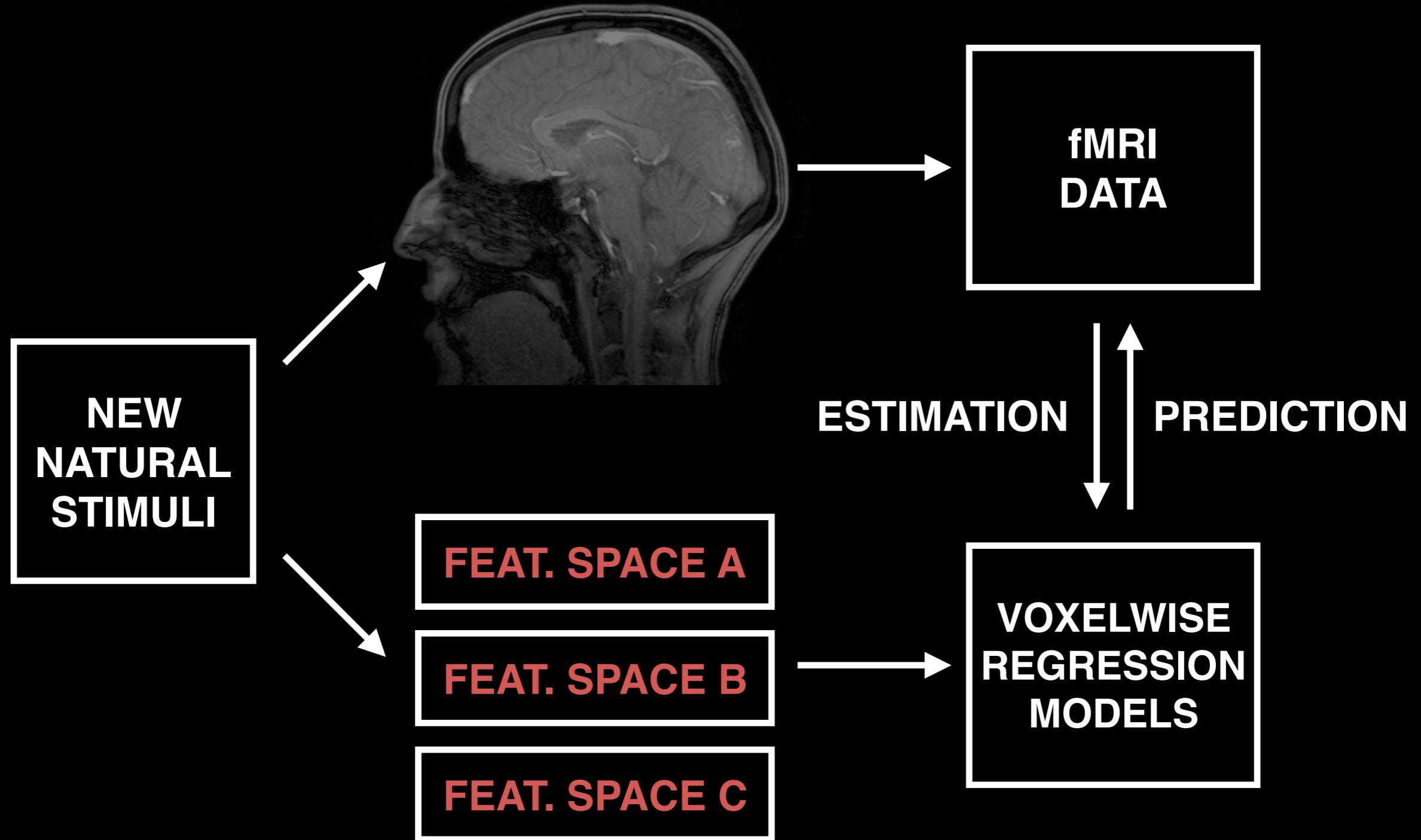
VOXELWISE MODELING



VOXELWISE MODELING



VOXELWISE MODELING



NATURAL LANGUAGE EXPERIMENT

- Subjects listen to over 2 hours of naturally spoken, narrative stories
- BOLD responses recorded from whole brain using GE-EPI fMRI
 - Voxel size: $2.24 \times 2.24 \times 4.1\text{mm}$
 - Repetition time (TR): 2.0045 s
 - Custom water-excitation RF pulse



Huth, de Heer, Griffiths, Theunissen, & Gallant. *Nature* (2016)

fMRI DATA VISUALIZATION



VOXELWISE MODELING

SIMPLEST MODEL:

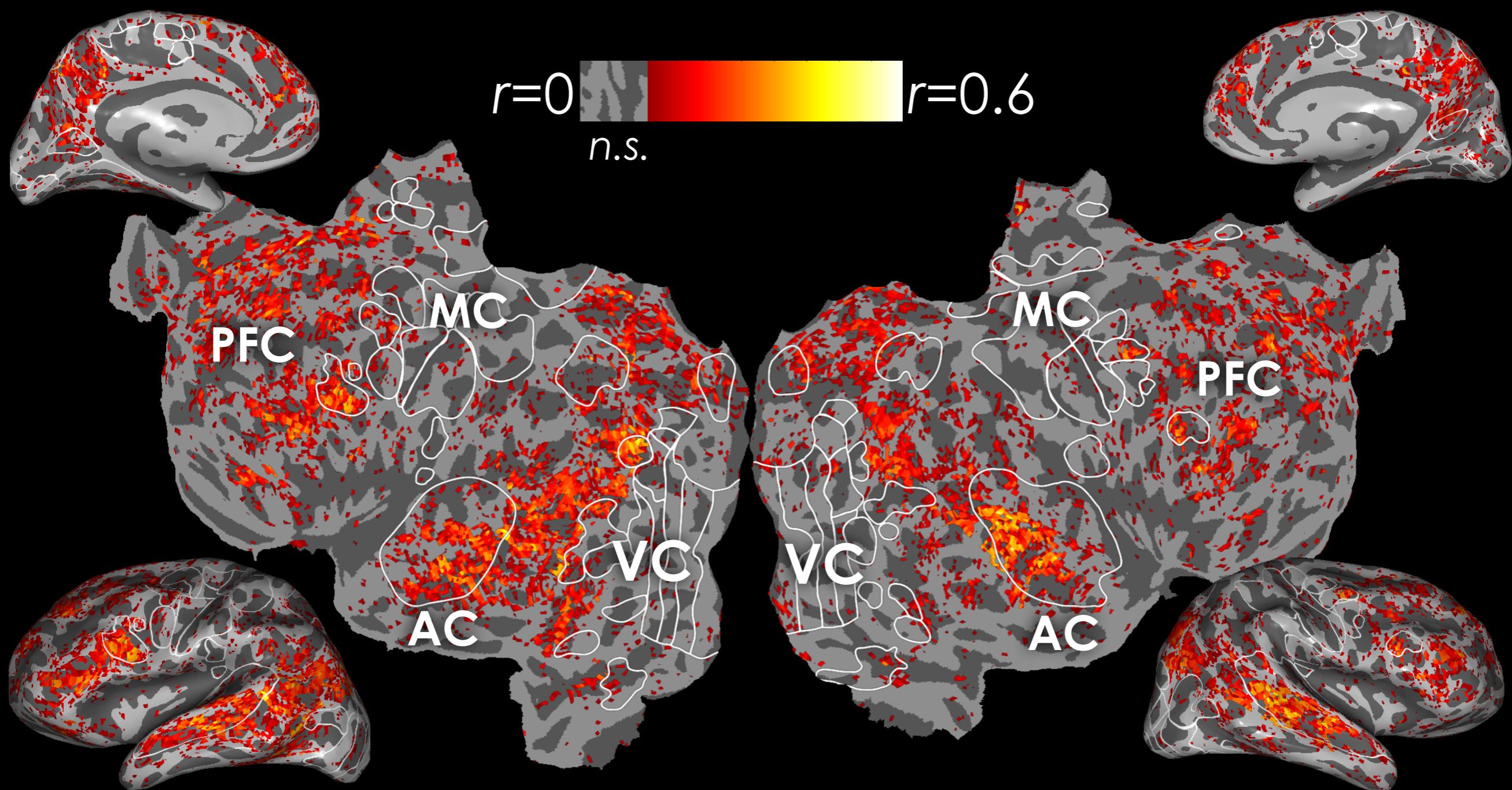
each voxel responds (some amount) to each word

$$R(t) = \sum_{i=0}^N \beta_i W_i(t)$$

$$\hat{\beta} = \operatorname{argmax}_{\beta} P(R|\beta, W)$$

likelihood

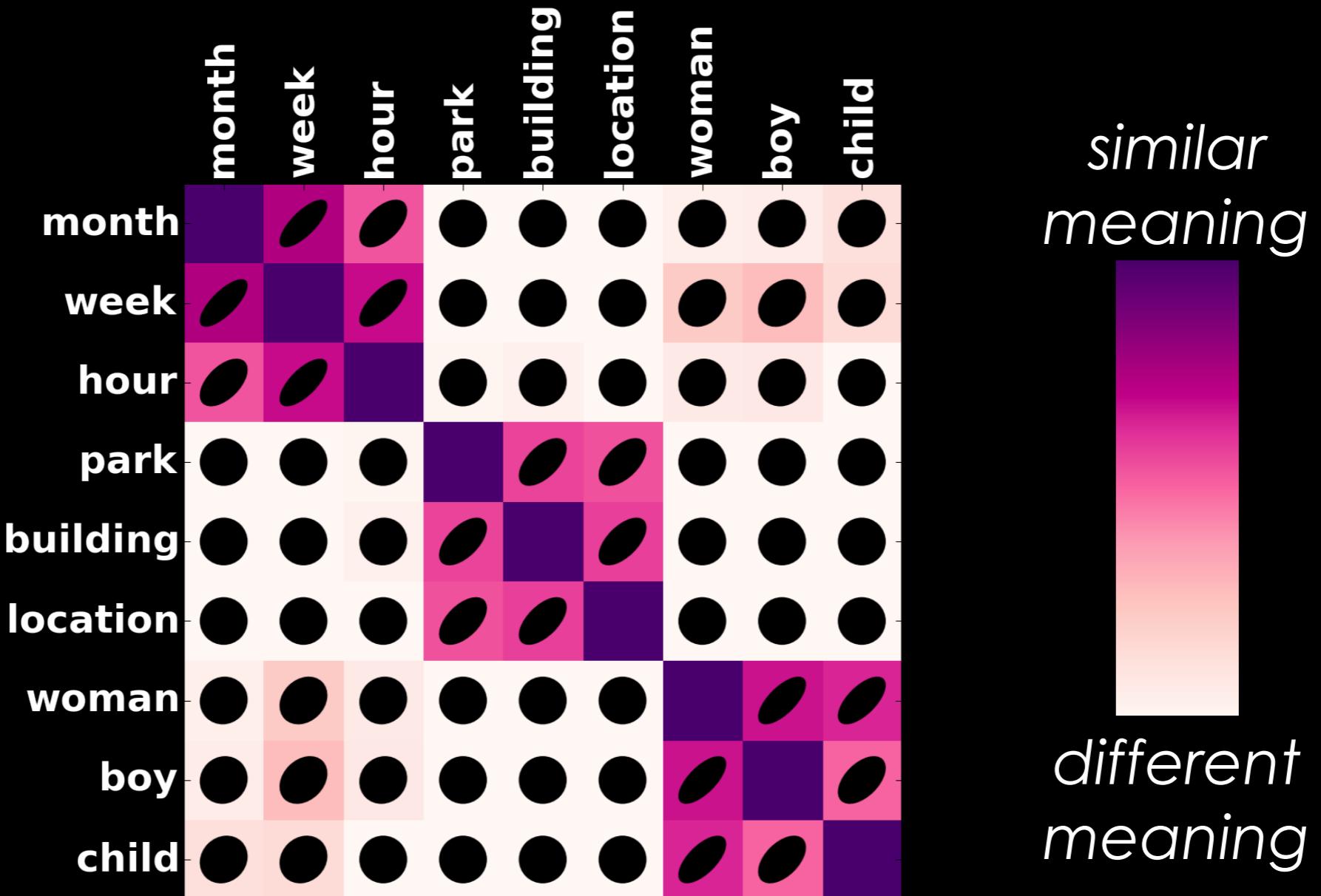
WORD MODEL PERFORMANCE: MEDIOCRE



SEMANTIC PRIOR

IMPROVED MODEL:

similar responses to words with similar meanings



SEMANTIC PRIOR

Distributional hypothesis:

“You shall know a word by
the company it keeps”

J. R. Firth (1954)

SEMANTIC PRIOR

...

difficult

...

husband

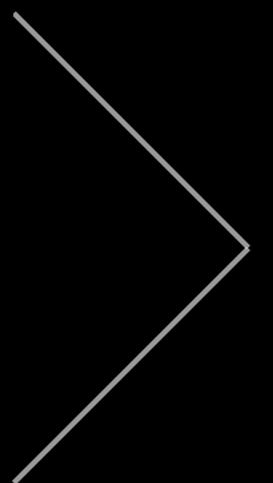
...

potato

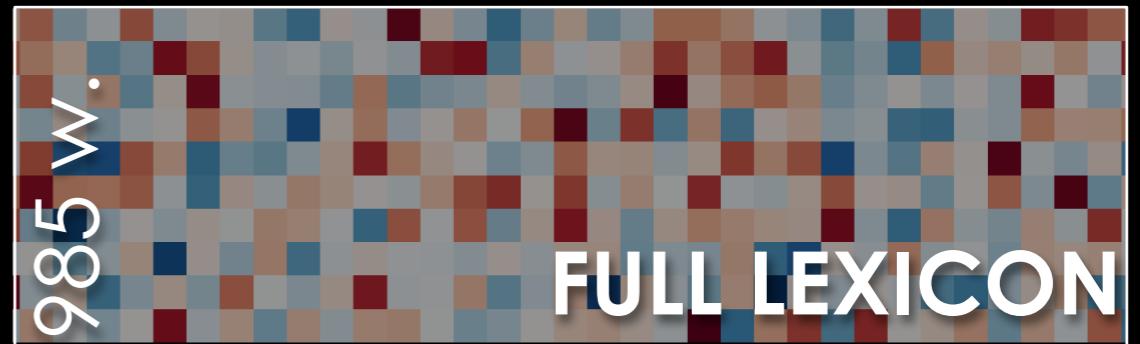
...

remember

...



TARGET
WORDS

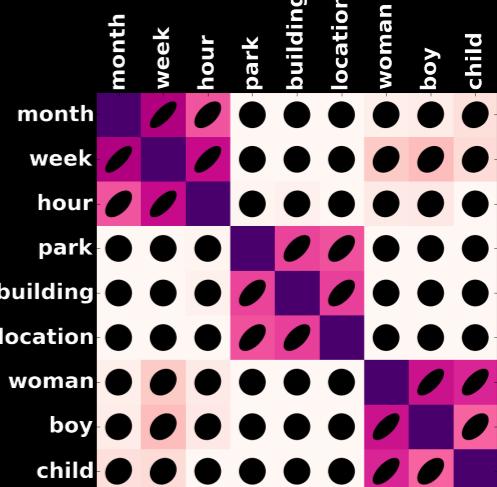


10,470 words

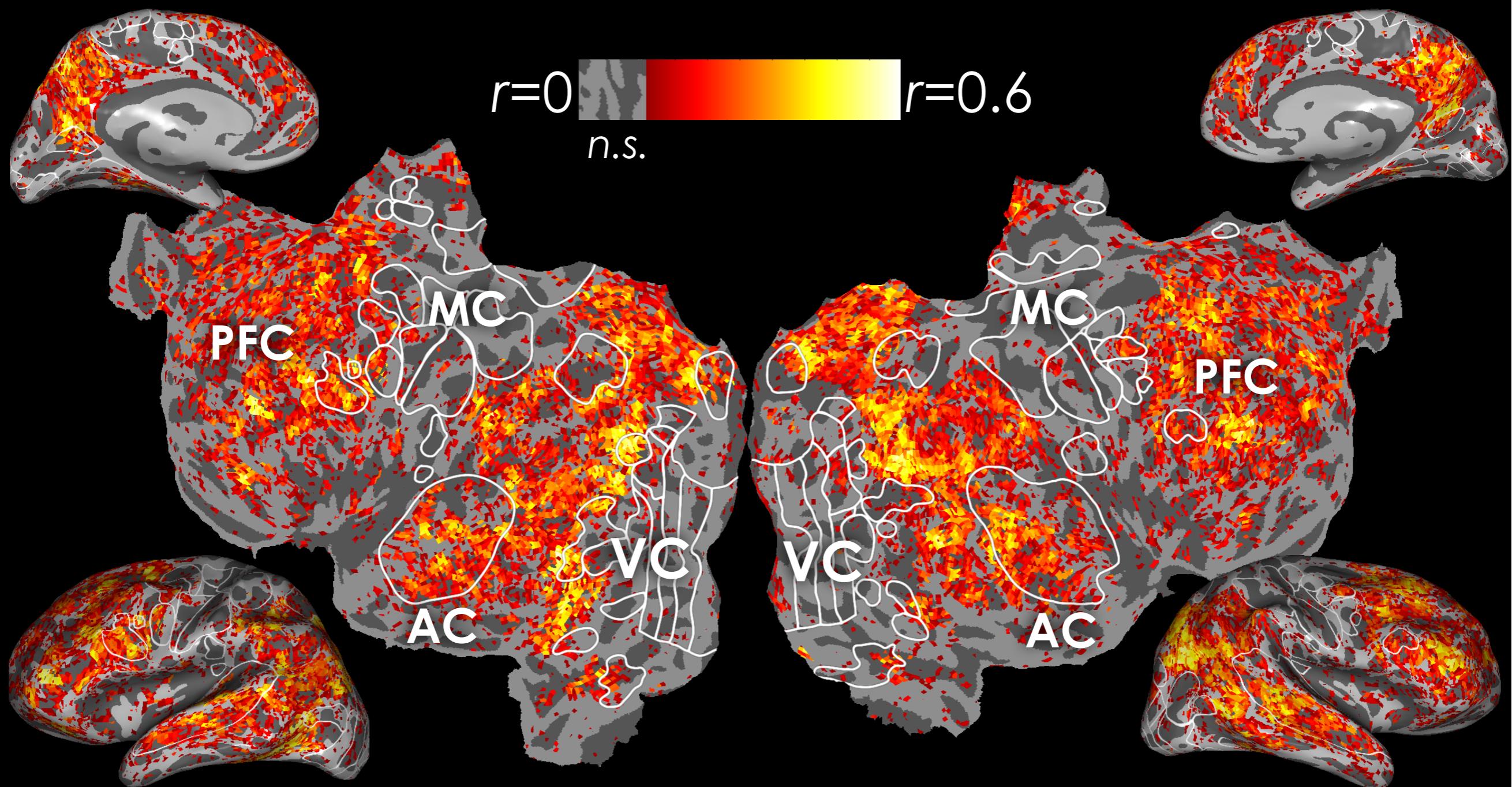
SEMANTIC PRIOR

IMPROVED MODEL:

similar responses to words with similar meanings



SEMANTIC MODEL PERFORMANCE: EXCELLENT



MODEL INTERPRETATION

visual
tactile
abstract
numeric

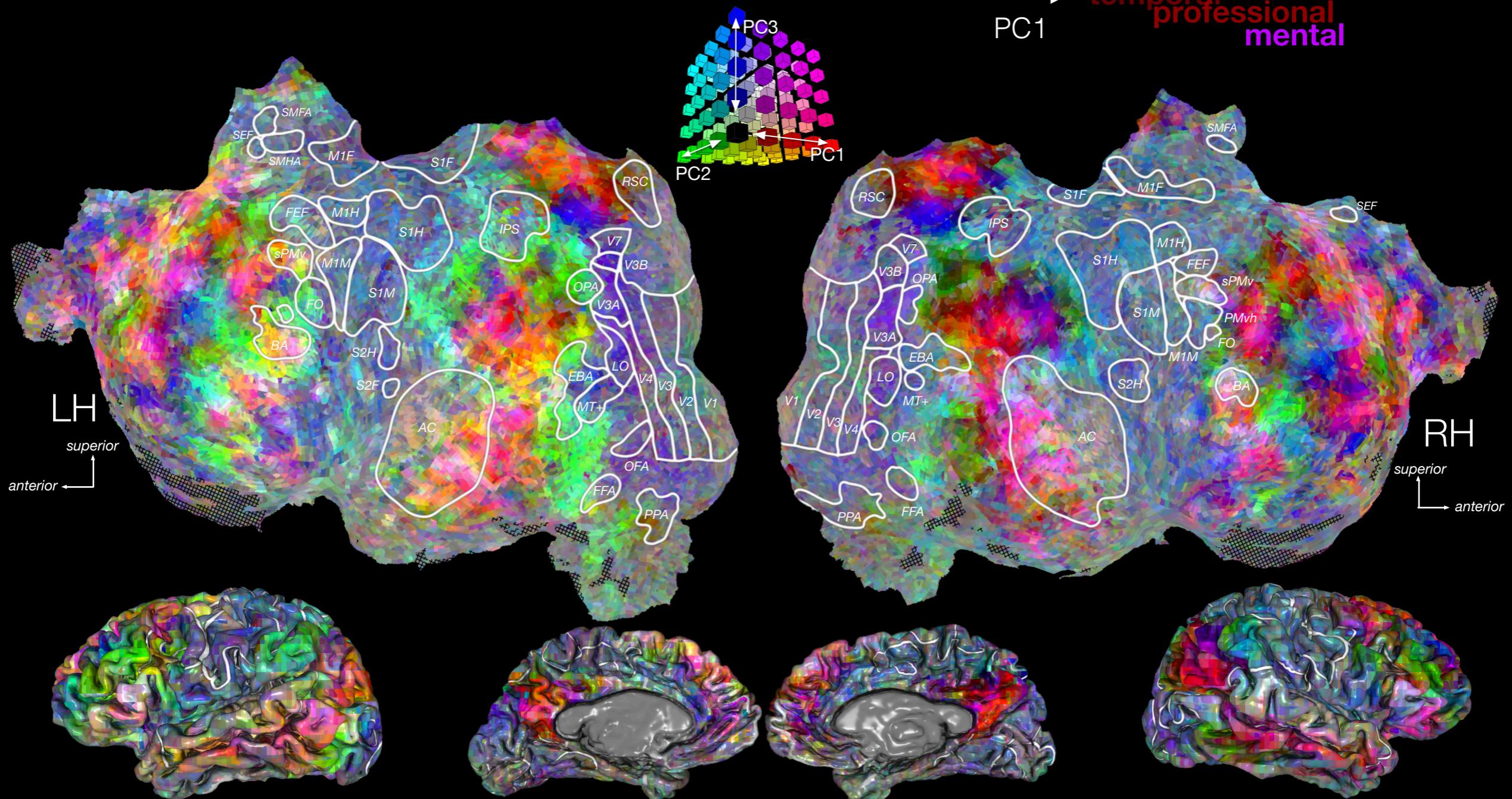
violent
communal
emotional
social

locational

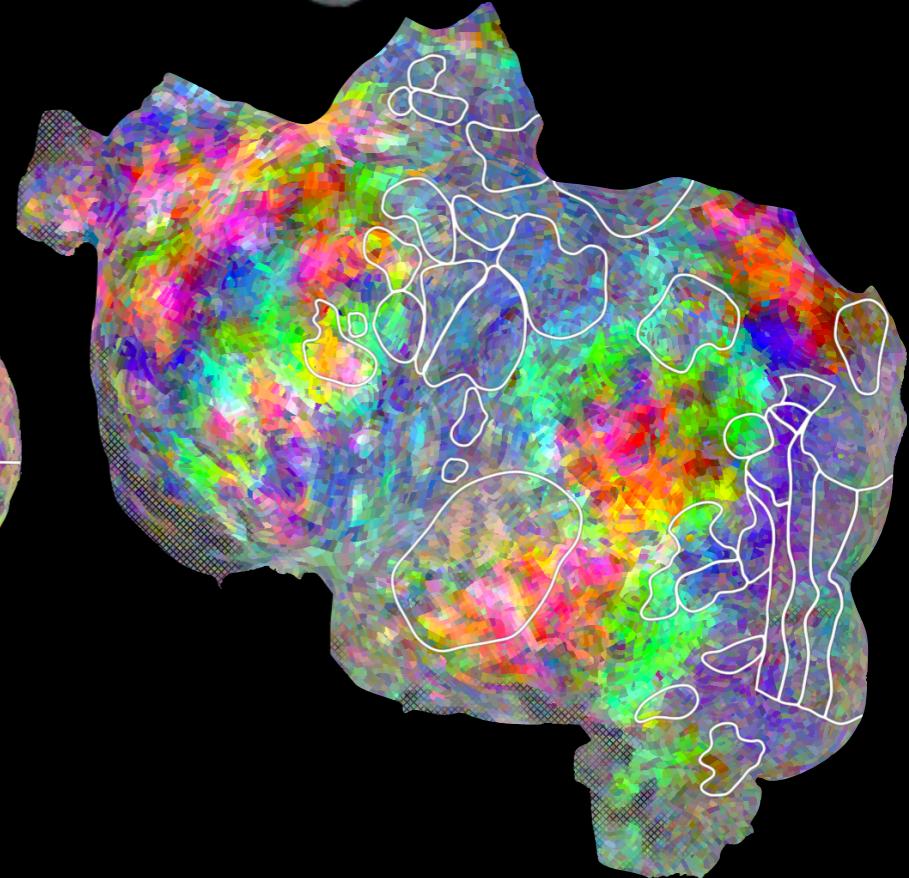
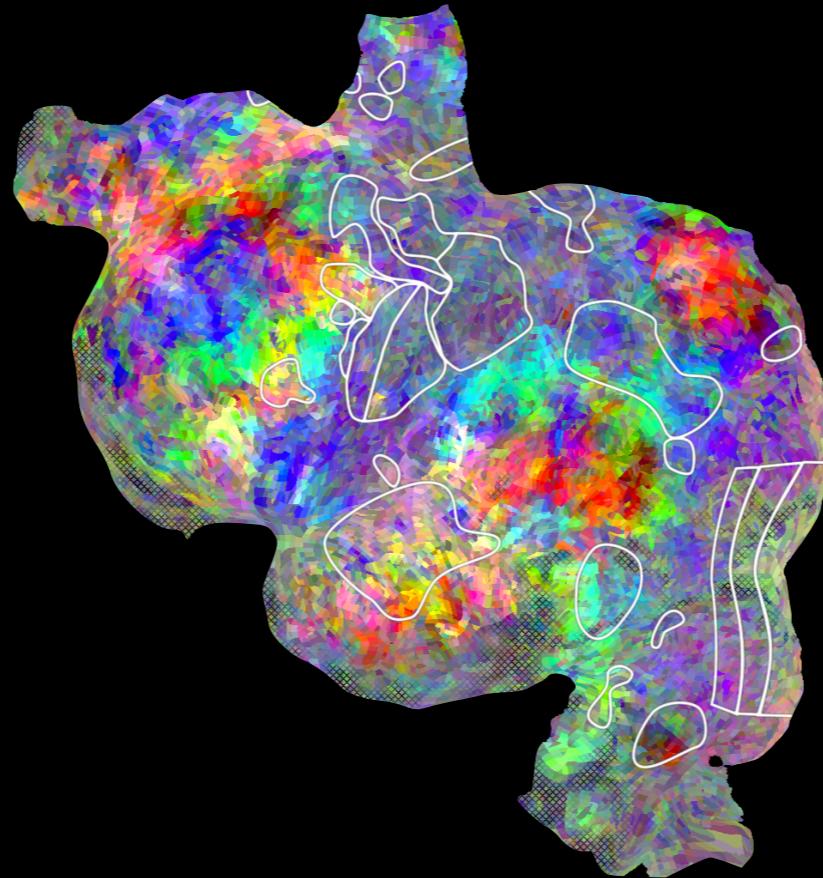
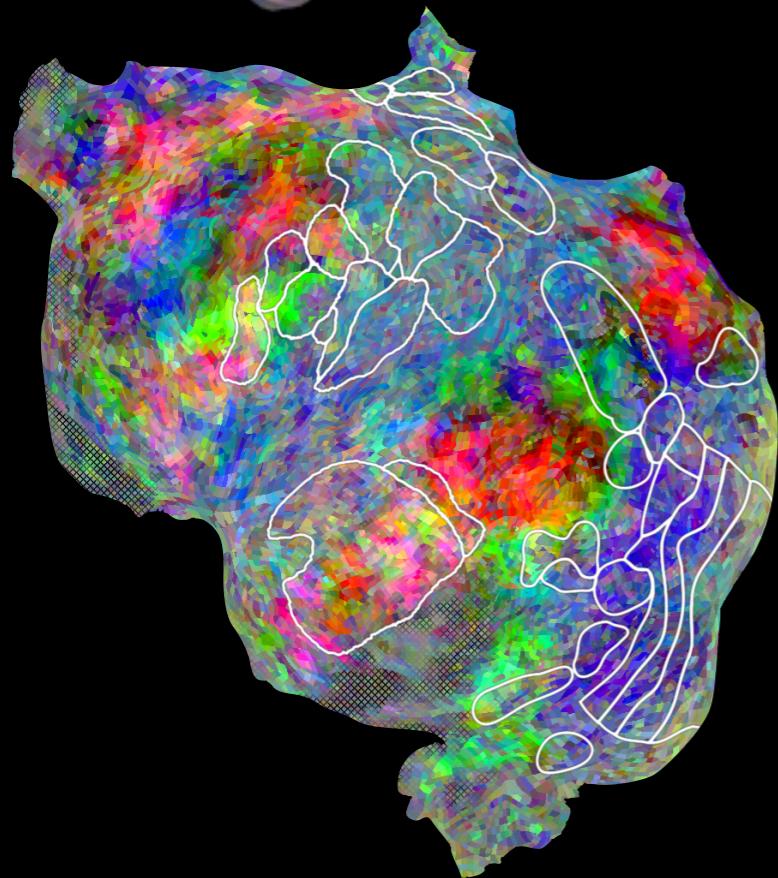
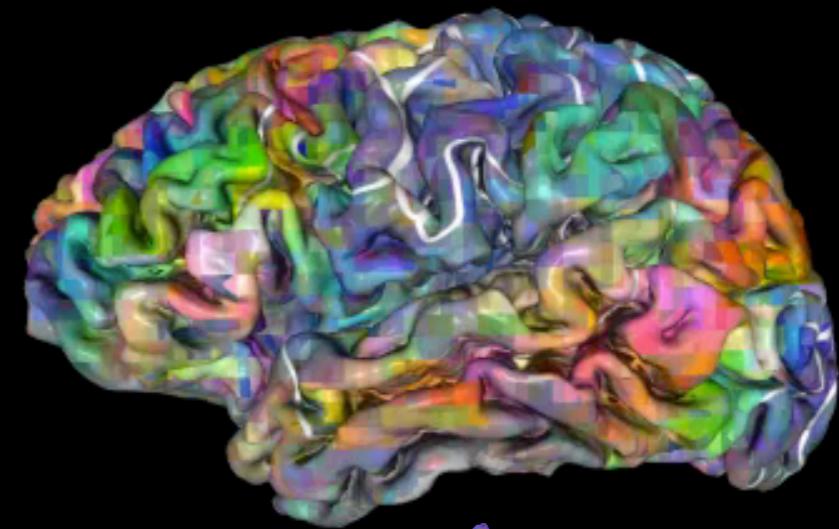
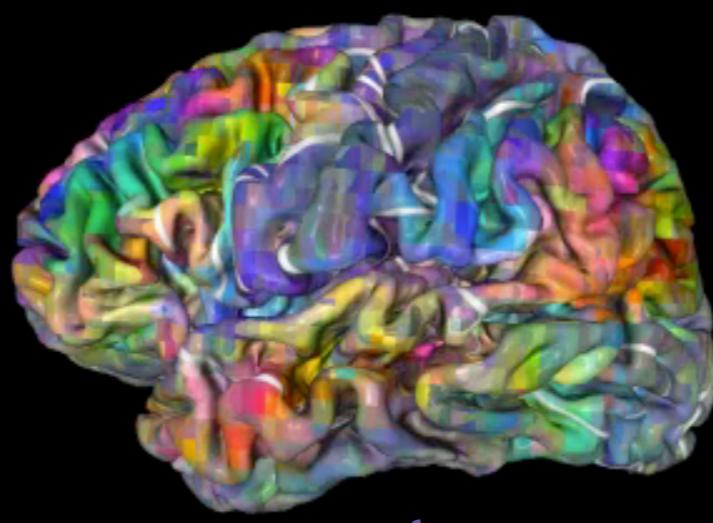
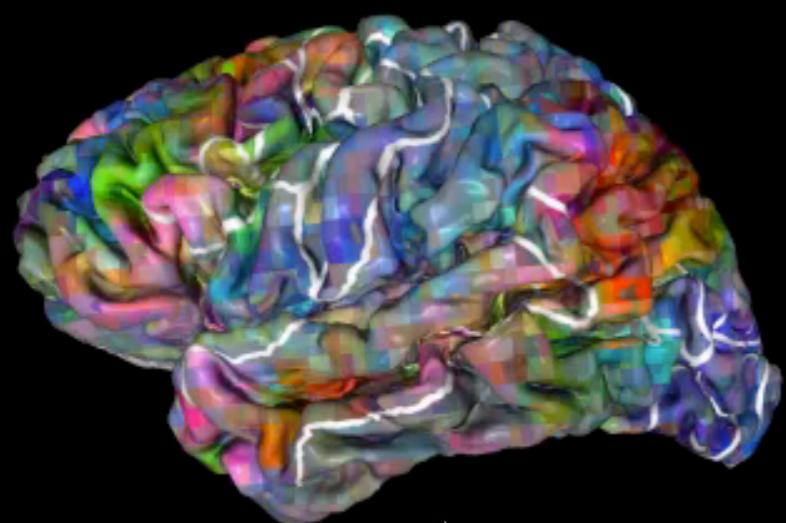
PC1

PC2

temporal
professional
mental

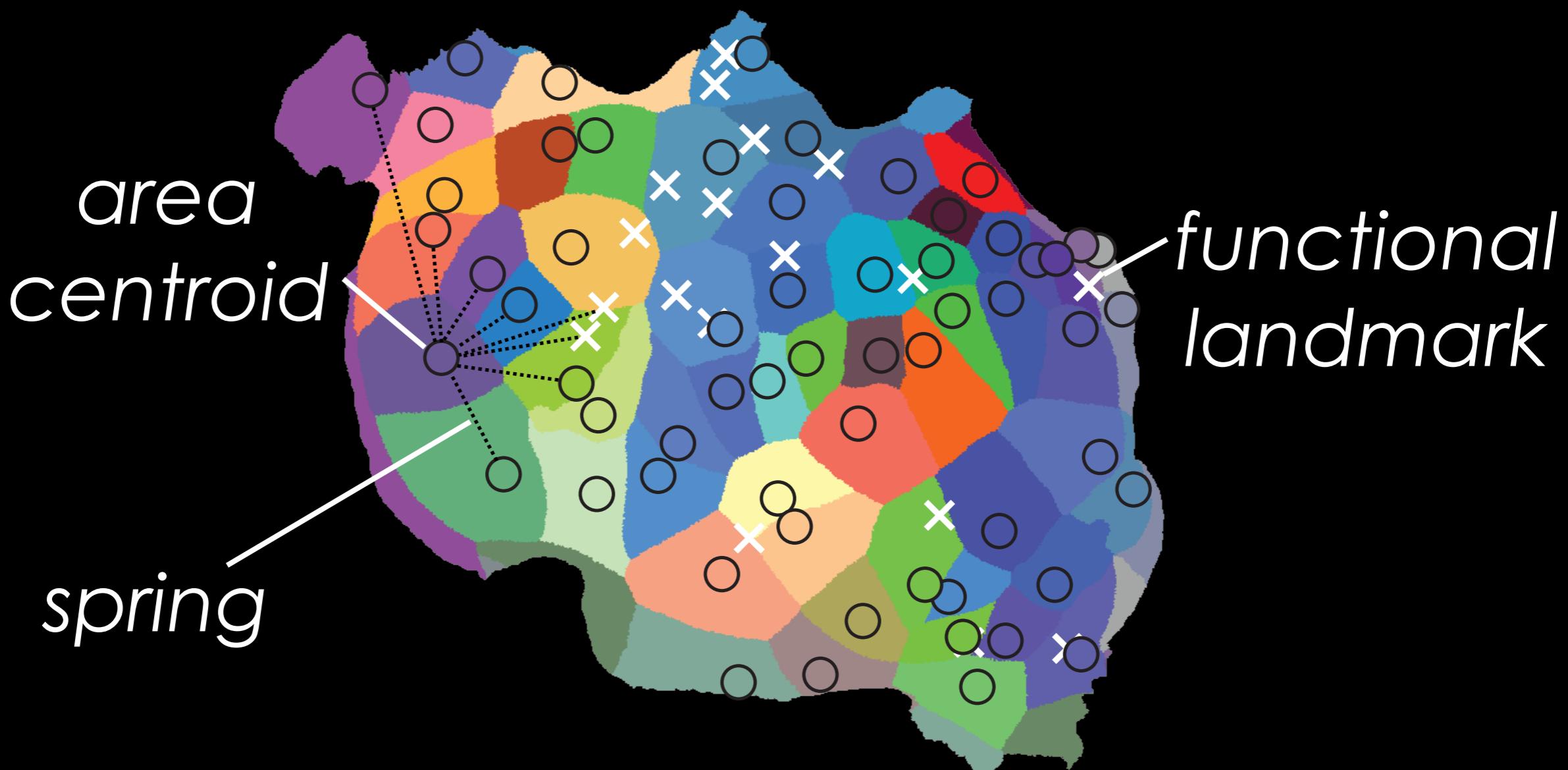


MAPS ARE CONSISTENT ACROSS SUBJECTS



PrAGMATIC

Probabilistic and Generative
Model of Areas Tiling the Cortex



PrAGMATIC

SHARED
ACROSS
SUBJECTS

Ideal Spring Lengths

Spring Constants

Area Semantic Properties

UNIQUE TO
EACH
SUBJECT

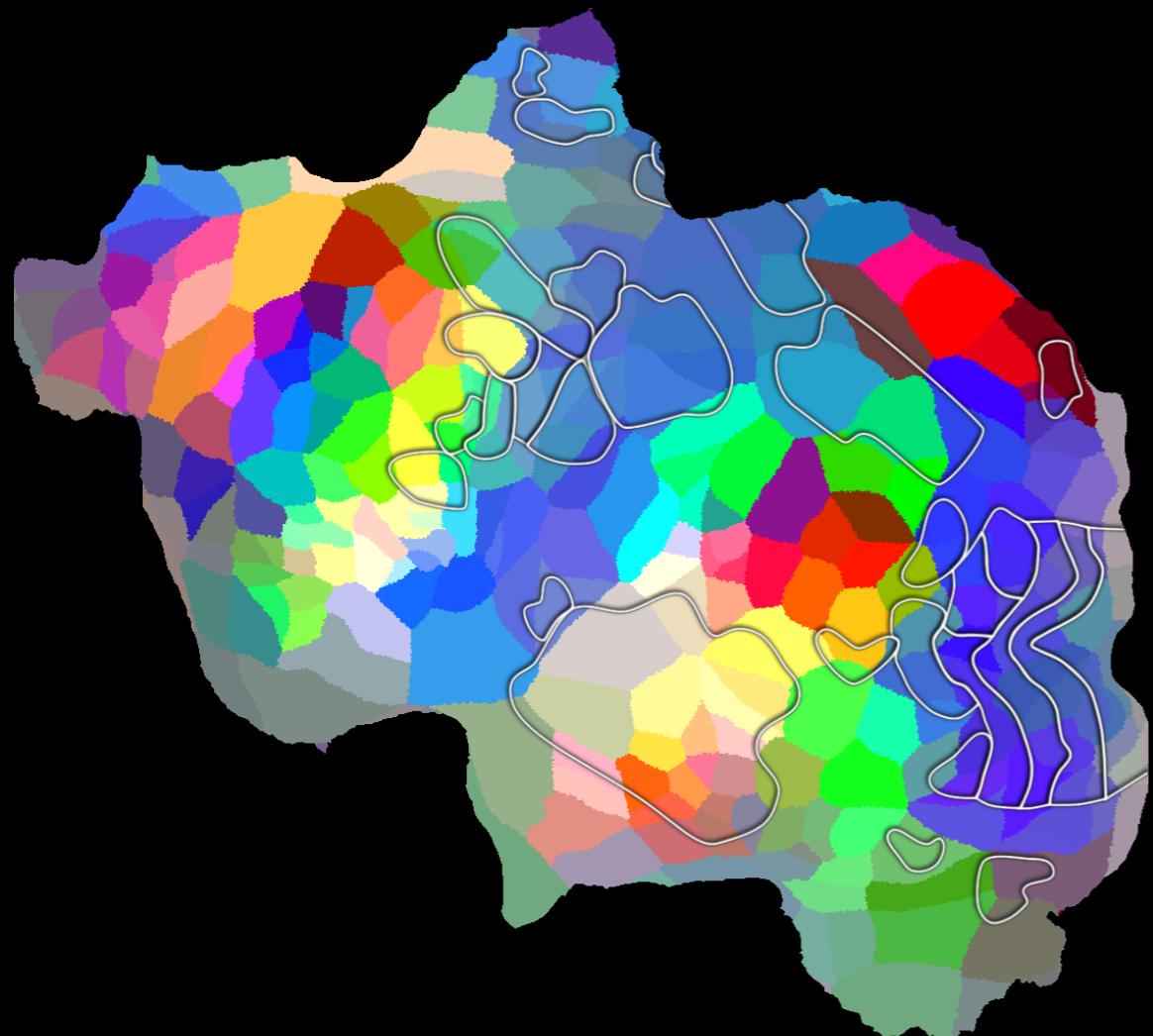
Exact Area Locations

Observed Functional Data

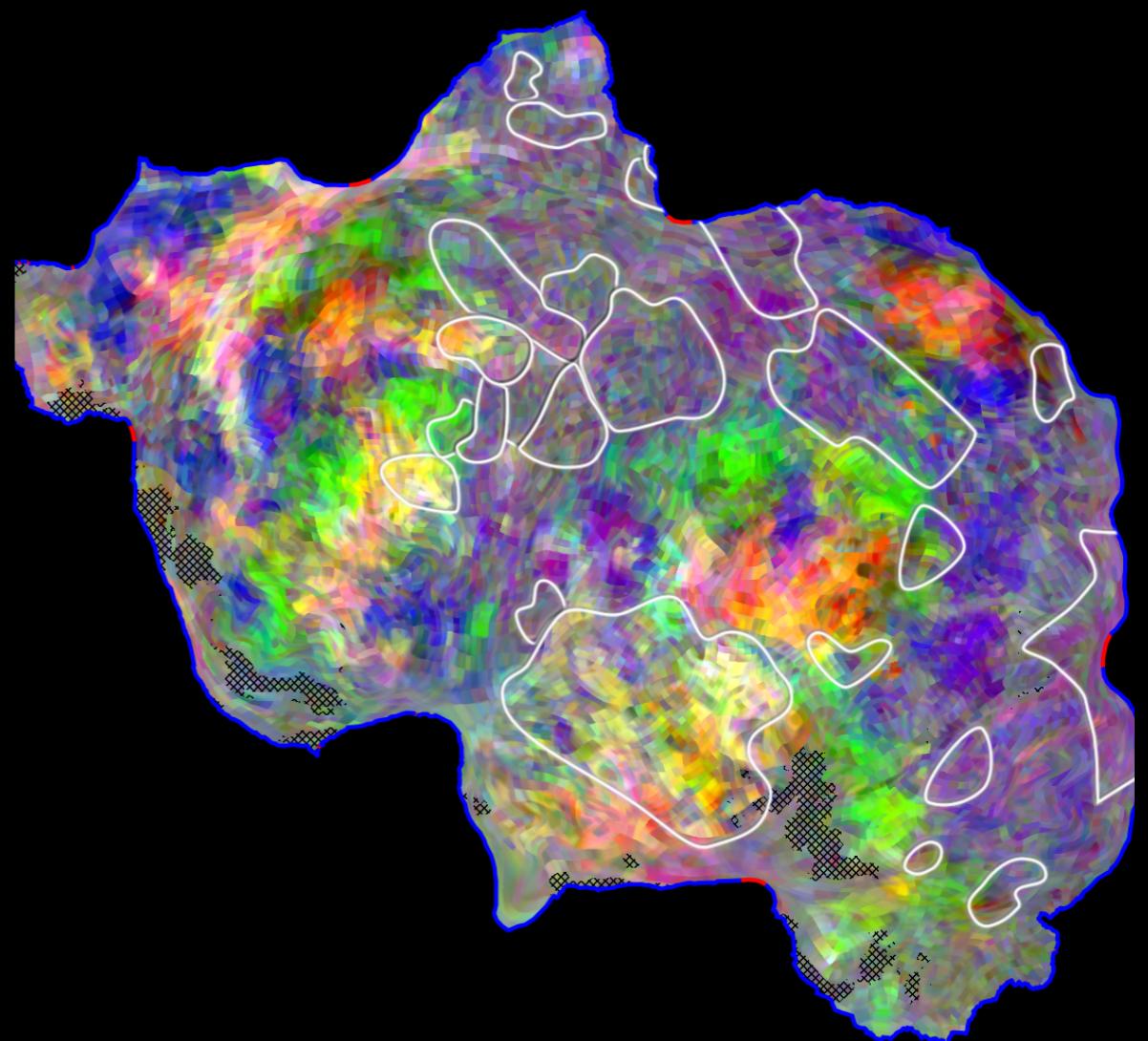
PrAGMATIC

model tested on held-out subjects

Predicted map



Actual map



PrAGMATiC Parameters

Shared across subjects

L - ideal spring lengths

K - spring constants

M - area functional means

Unique to each subject

H - exact centroid locations

V - functional values on cortex

$$P(V, H; M, L, K) = P(H; L, K) \quad P(V|H; M)$$

total probability

*arrangement
probability*

*map
probability*

$$P(H; L, K) \propto e^{-E(H; L, K)}$$

$$E(H; L, K) = \sum_{i,j,s} K_{ij} (d_{ijs} - L_{ij})^2$$

distance between centroids i and j in subject s

$$P(V|H; M) \propto e^{-E(V|H; M)}$$

$$E(V|H; M) \propto \sum_{l,s} (M_{H(s,l)} - V_{ls})^2$$

nearest centroid in subject s

SYLLABUS

- * Online:
github.com/alexhuth/n4cs-fa2017/wiki
- * 2 problem sets (due Oct. 12 & Oct. 26)
- * Final project (alone or in pairs)
 - * Proposal due Oct. 31
 - * In-class presentations Dec. 5 & 7
 - * Write-up due Dec. 5

THAT'S IT