

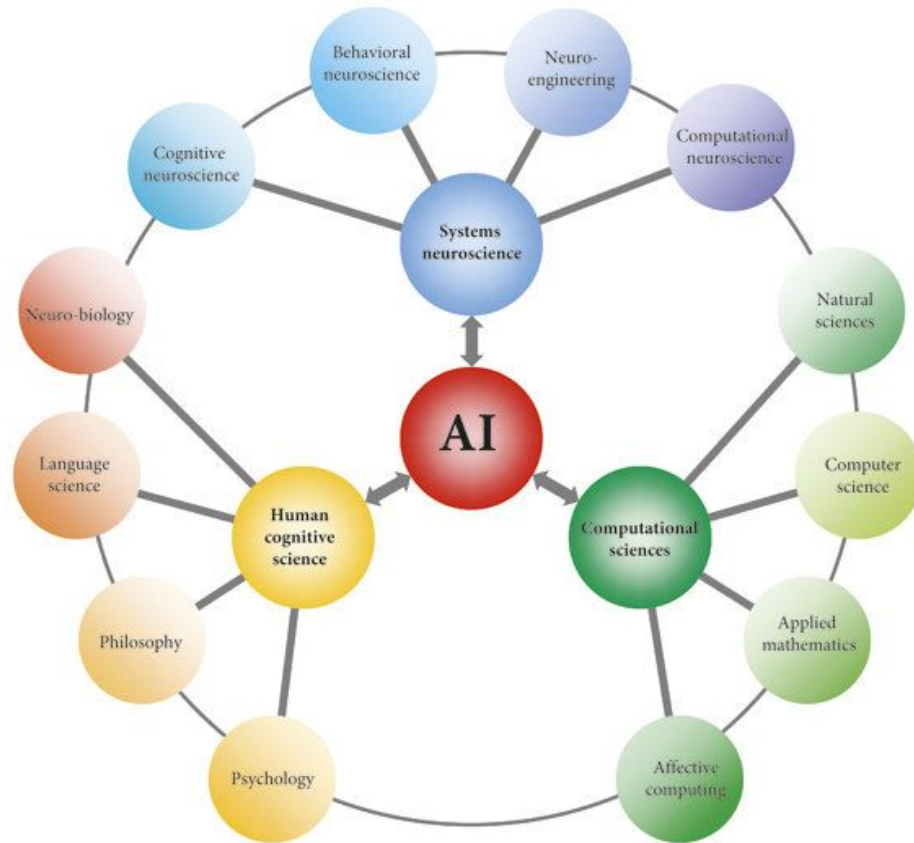
Applications of AI: Cognitive Science

Naiti Bhatt
University of Edinburgh

Roadmap

- Cognitive Science
- Measuring cognition: Neuroimaging
- Algorithmic tools for modeling cognitive systems

Cognitive Science



What is the difference between
the brain and the mind?

Brain versus Mind

Brain

Anatomical Organ

Vehicle for thoughts

Composed of neurons

Involuntary operator of consciousness

Hardware of the human computer

Mind

Hypothetical Construct

Origin of thoughts

Composed of phenomena

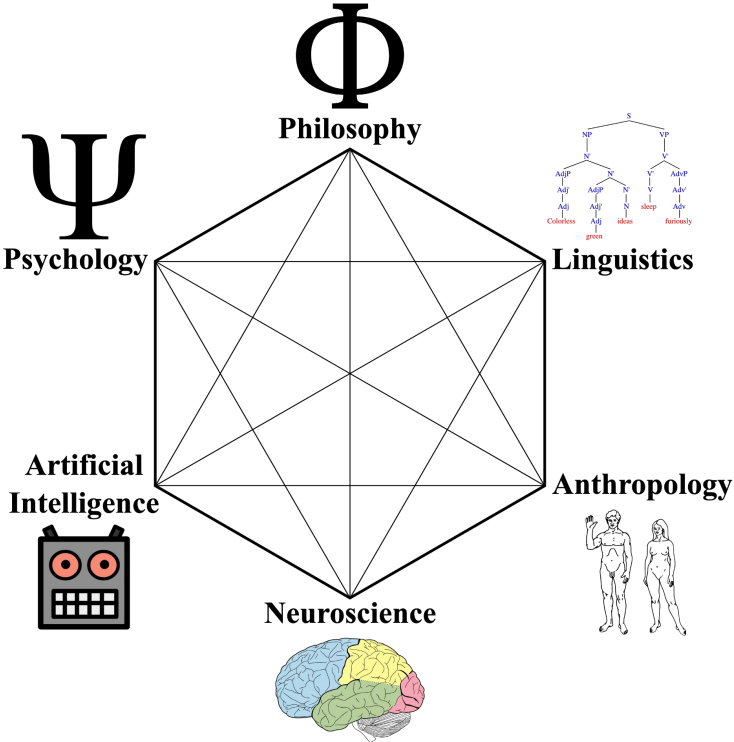
Volitional source of consciousness

Software of the human computer

What is cognitive science?

The study of the human mind and brain, focusing on how **the mind** represents and manipulates **knowledge** and how mental representations and processes are realized in **the brain**.

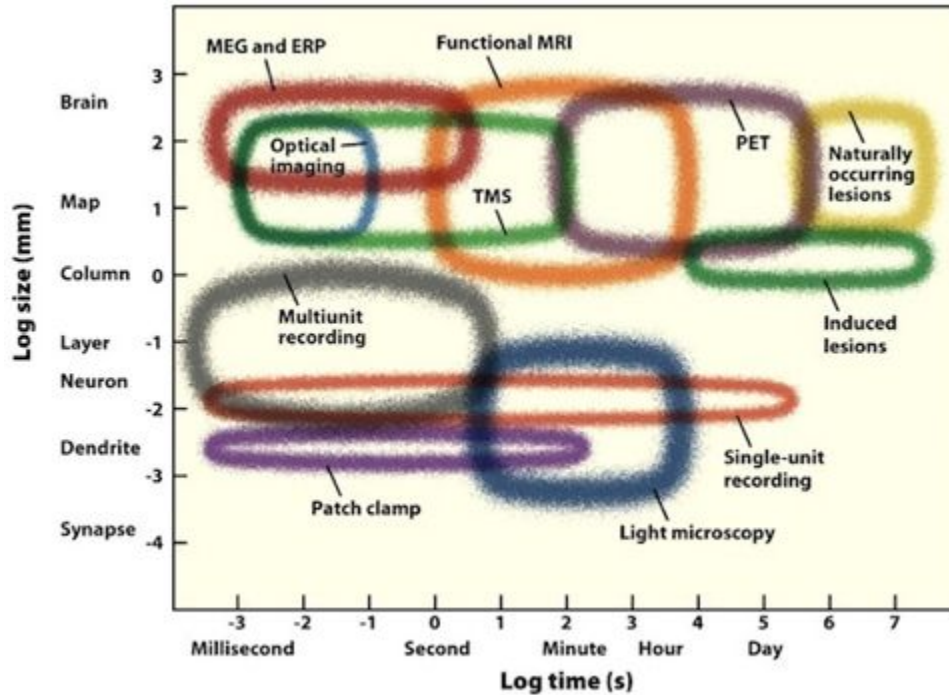
Quintessentially Interdisciplinary!



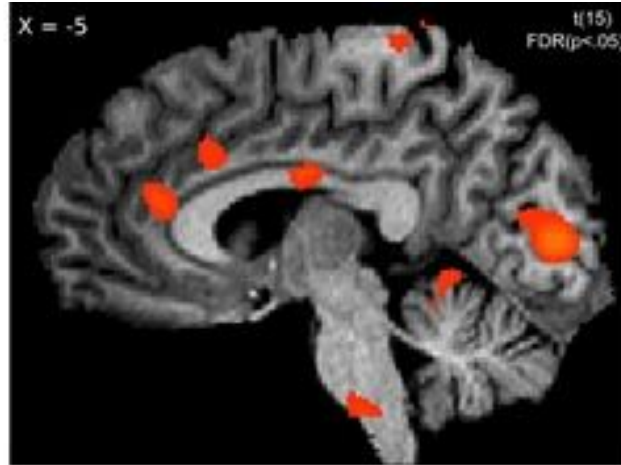
How do we do
research in
cognitive science?

Measuring cognition: Neuroimaging

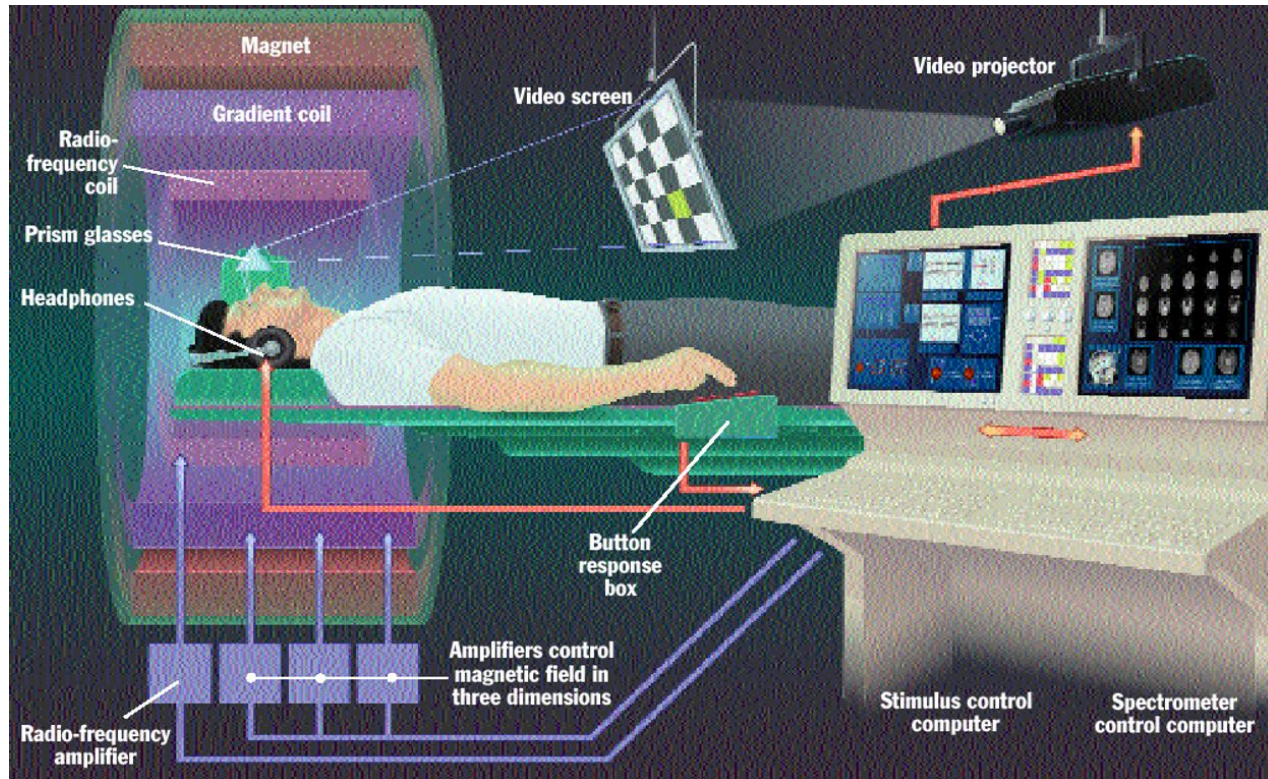
Loads of different neuroimaging methods!



Magnetic Resonance Imaging

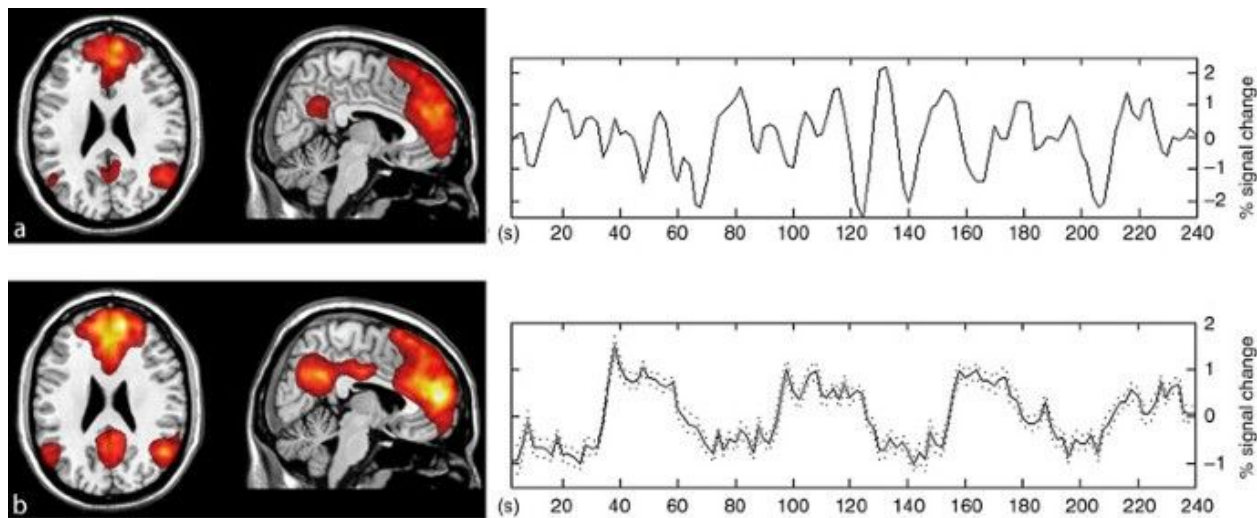


fMRI Setup



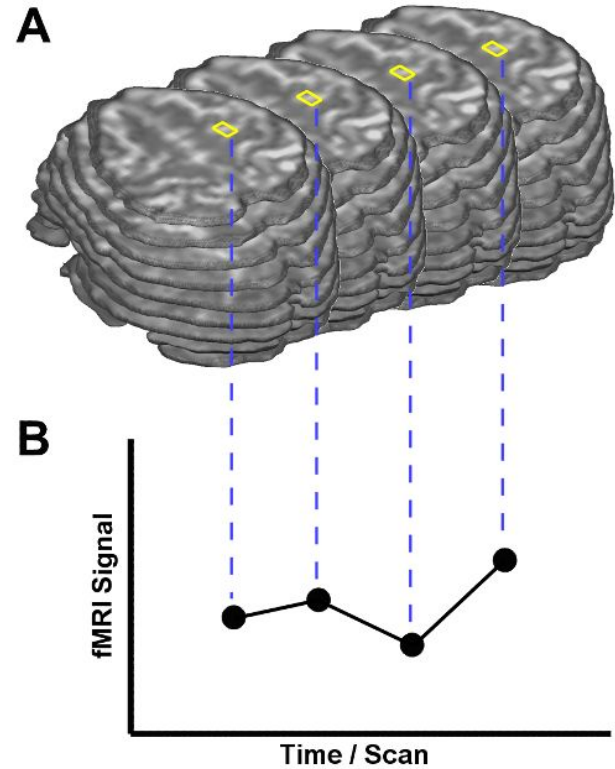
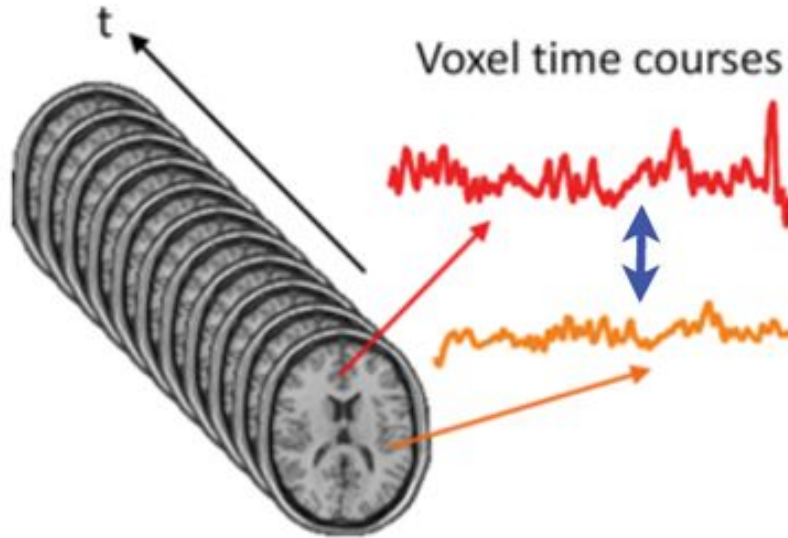
fMRI: Functional Magnetic Resonance Imaging

- Noninvasive and indirect method of measuring neural 'activity'
- Uses flow of oxygenated blood as a proxy
- Measure voxel-wise



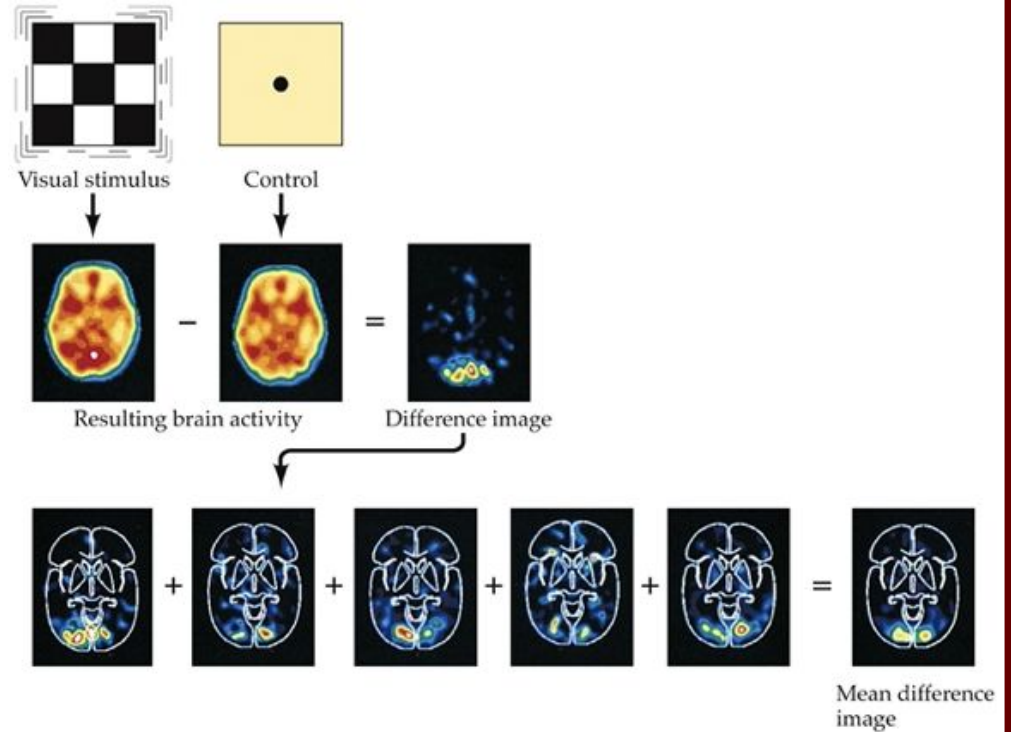
Voxel (**V**olumetric **P**ixel)

Usually, $2\text{-}3\text{mm}^3$

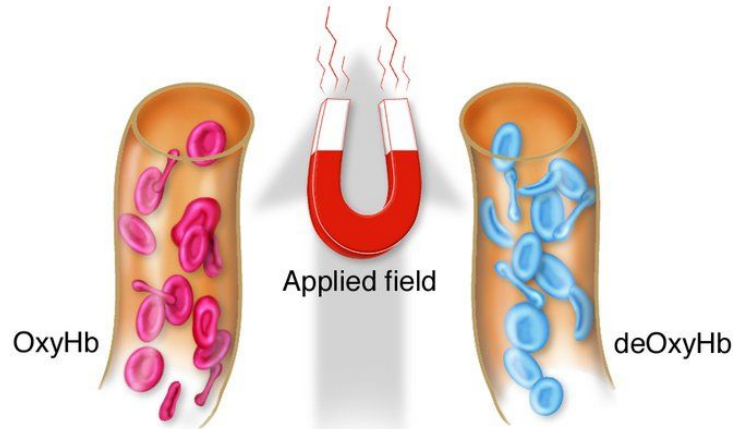


Final result: Activation Maps

Stimulus - Control = Activation



BOLD (Blood Oxygen Level Dependent) Response

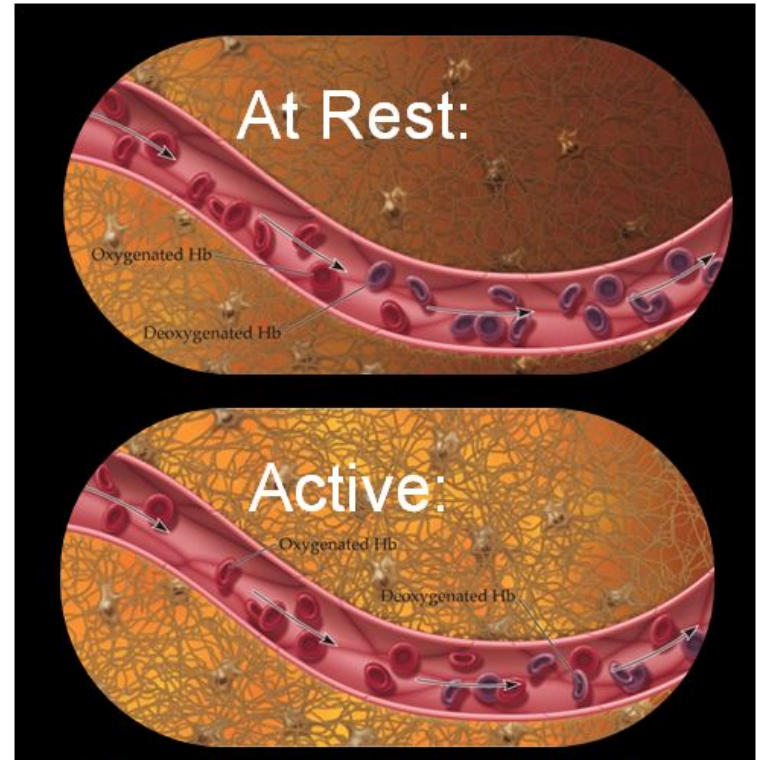


Magnetic susceptibility of OxyHb \approx tissue
Field uniform
MFI signal high

Magnetic susceptibility of deOxyHb \gg tissue
Field non-uniform
MRI signal low

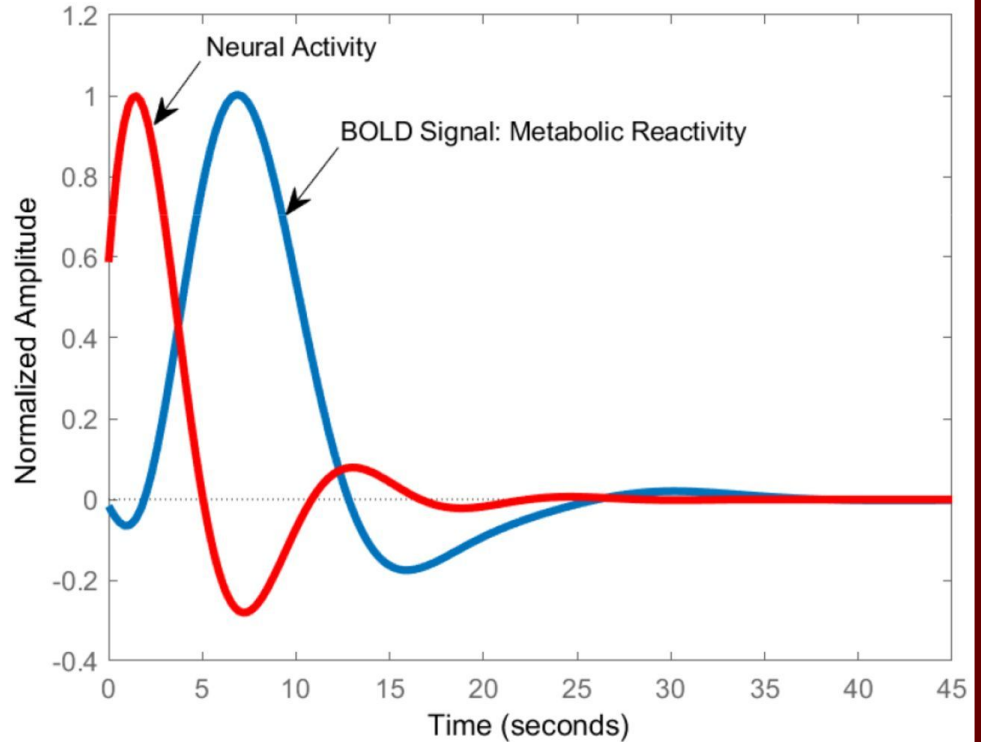


MRI signal

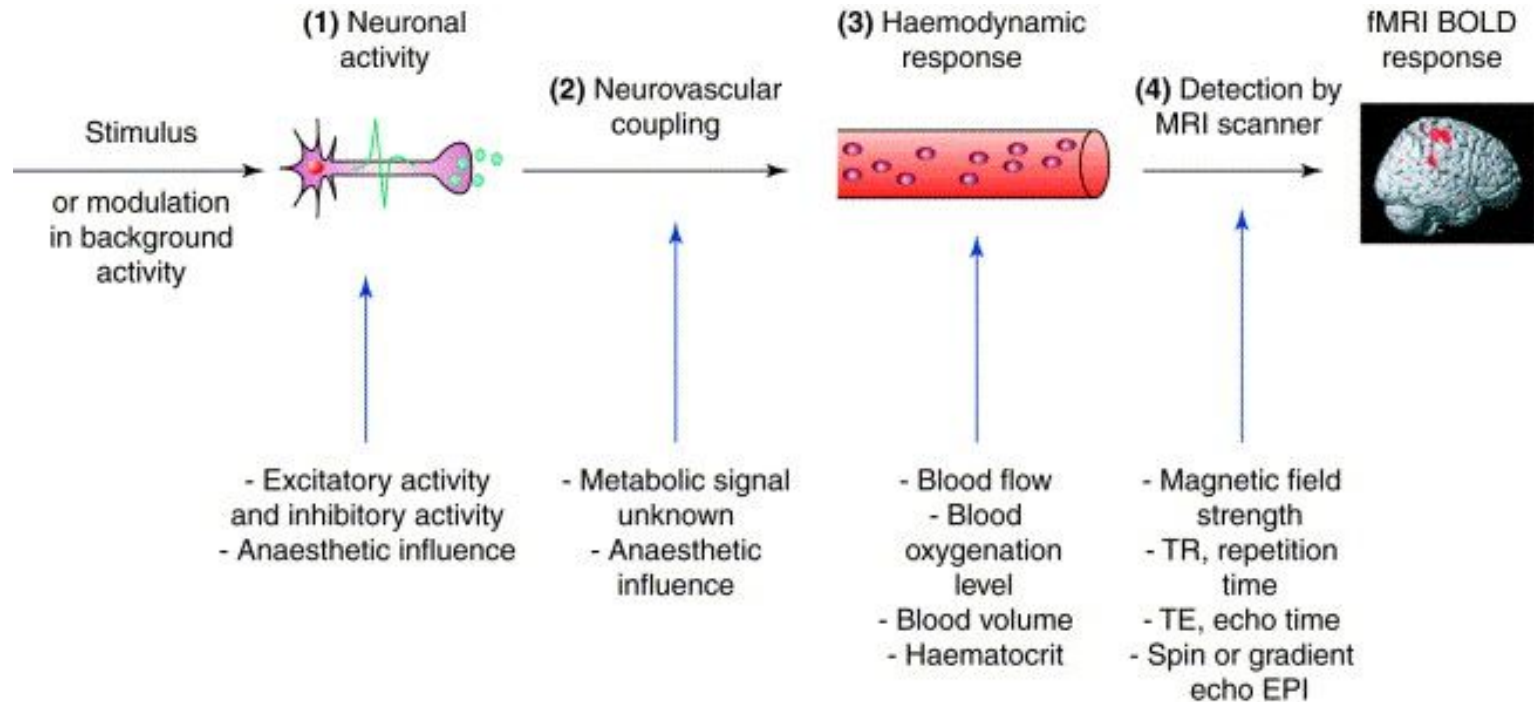


Hemodynamic Response Function (HRF)

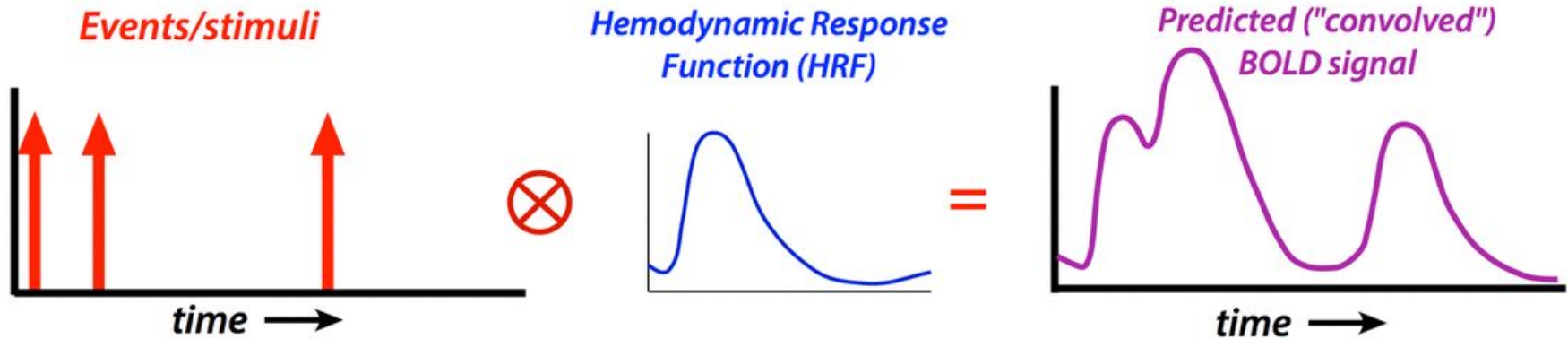
- Model of BOLD response
- Oxygenated blood follows neural activity
- Slower response gives time for measurement



Overview of the fMRI BOLD Signal



Predicting responses using convolution



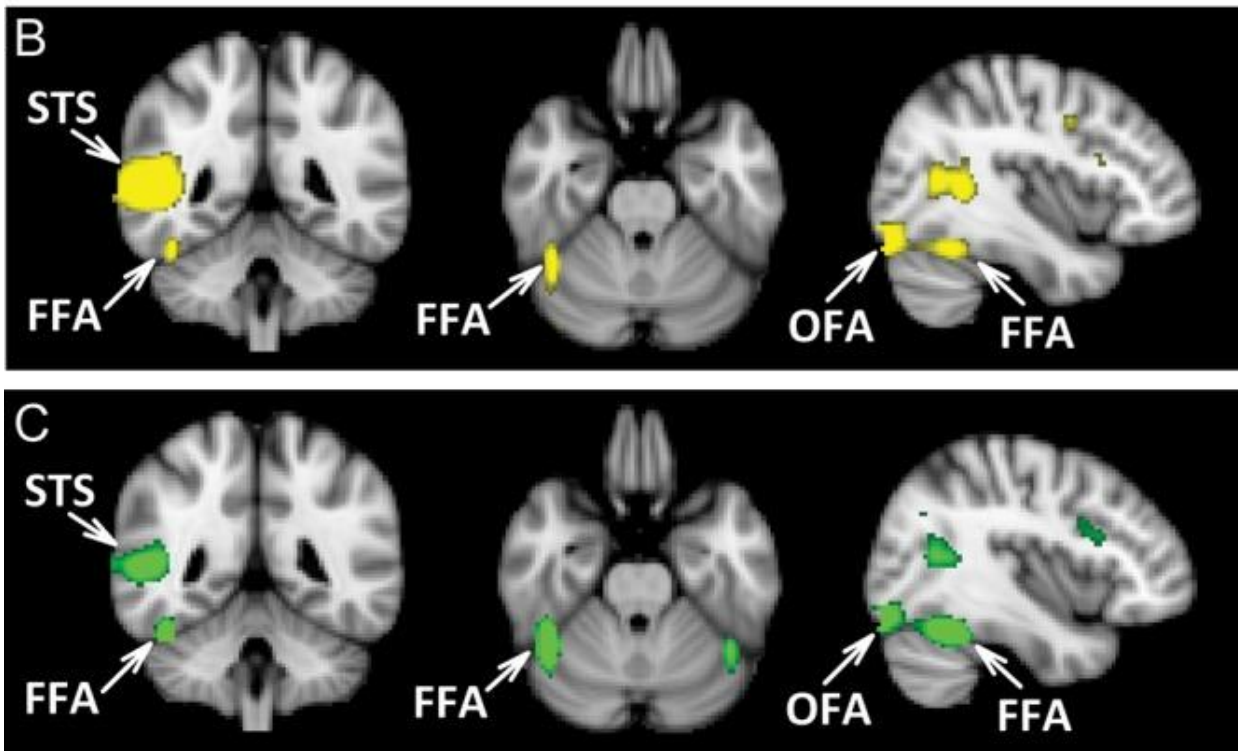
Two approaches to analyzing fMRI data

1. “Whole-brain” or “voxelwise” approach
2. Region-of-interest approach

1. “Whole-brain” or “voxelwise” approach

- For every voxel in the scanned volume (typically but not necessarily the whole brain), do a statistical test to look for expected differences in activation
 - e.g., ANOVA
 - main effect of stimulus category (faces vs. hands)
 - main effect of direction (left, direct, right)
 - interaction

1. “Whole-brain” or “voxelwise” approach

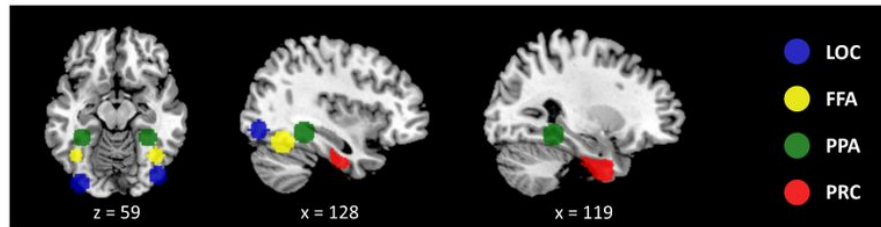


2. Region-of-interest approach

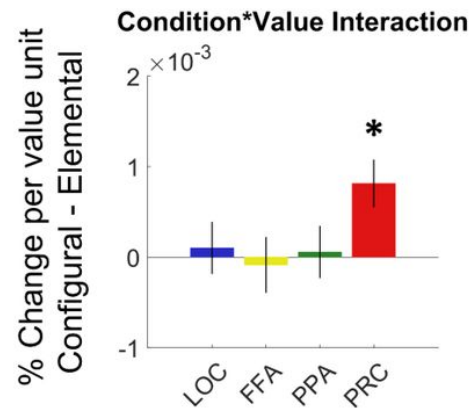
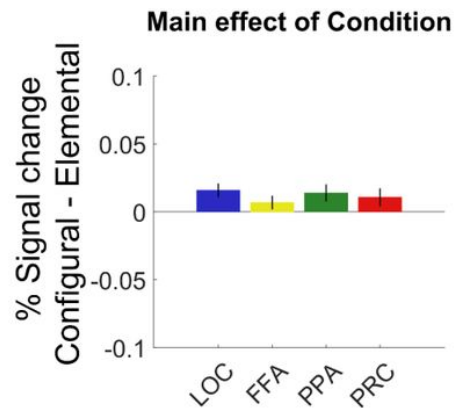
- Conduct separate “localizer” scan(s) to identify expected brain regions based on typical activation differences
 - e.g., Define FFA by a contrast of faces vs. other visual stimuli
- Regions can be identified in individual participants or across a group average
- Activation levels for different conditions can be extracted from main experiment data and tested

2. Region-of-interest approach

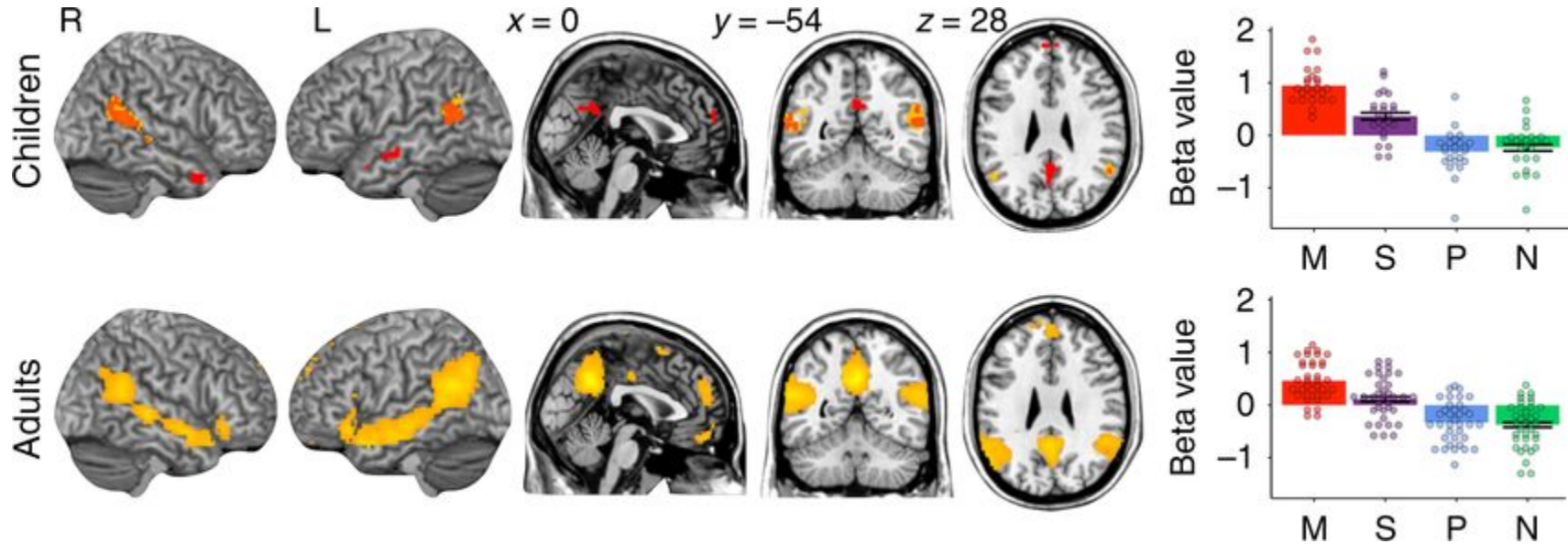
A. Ventral visual stream ROIs



B. Object presentation



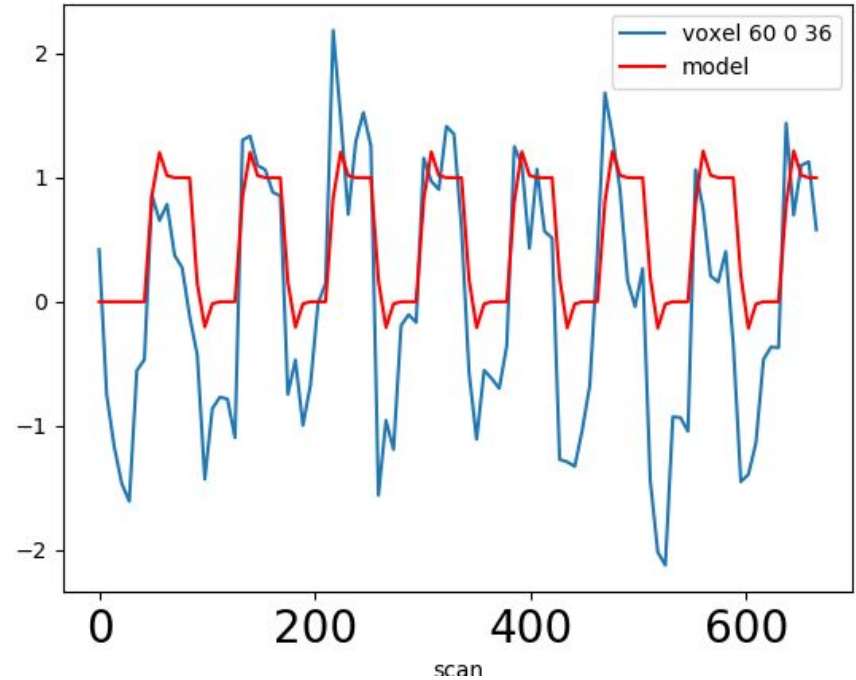
Using fMRI to study theory of mind development



fMRI Summary

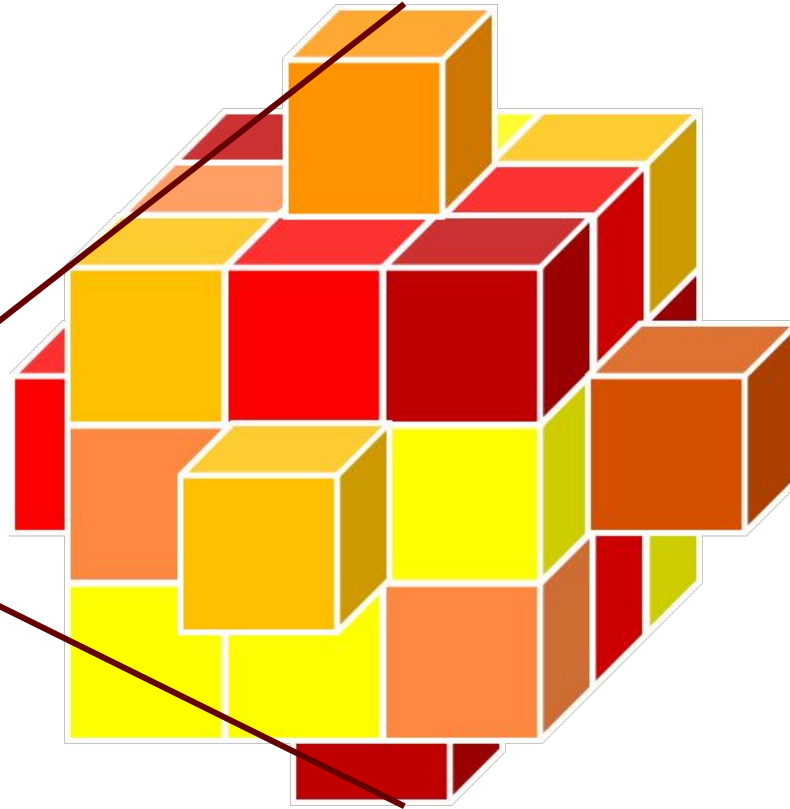
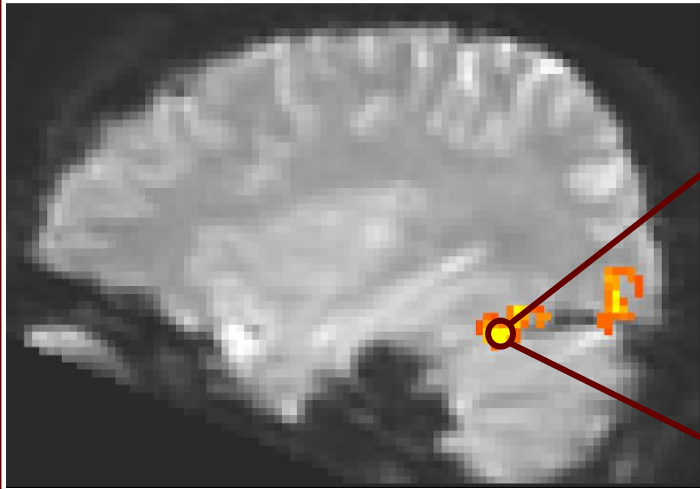
Measuring neural responses to
external stimuli

These data are really
multidimensional!



Algorithmic modeling tools

Region Of Interest (ROI): group of voxels



High activity



Low activity

Methods of Analysis

Univariate Analysis

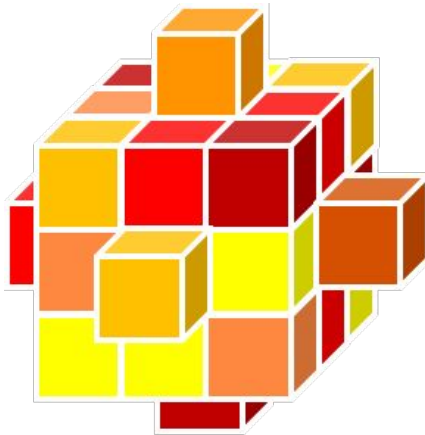
- look at one variable
- level of activation across a region
- average data from each voxel
- works well for strong differences

Multivariate Analysis

- look at multiple variables
- pattern of activation within a region
- preserve data from each voxel
- works well for weak effects across many voxels

Voxel Pattern Information

Condition 1
e.g., Faces



Condition 2
e.g., Hands



Two regions might have similar average **levels** of activation but different spatial **patterns** of activation

TRAINING

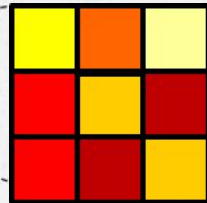
Image



fMRI scan

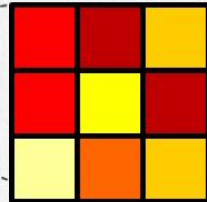


Voxel pattern



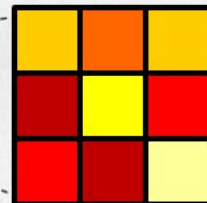
Output

=SHOE



=CAT

TESTING

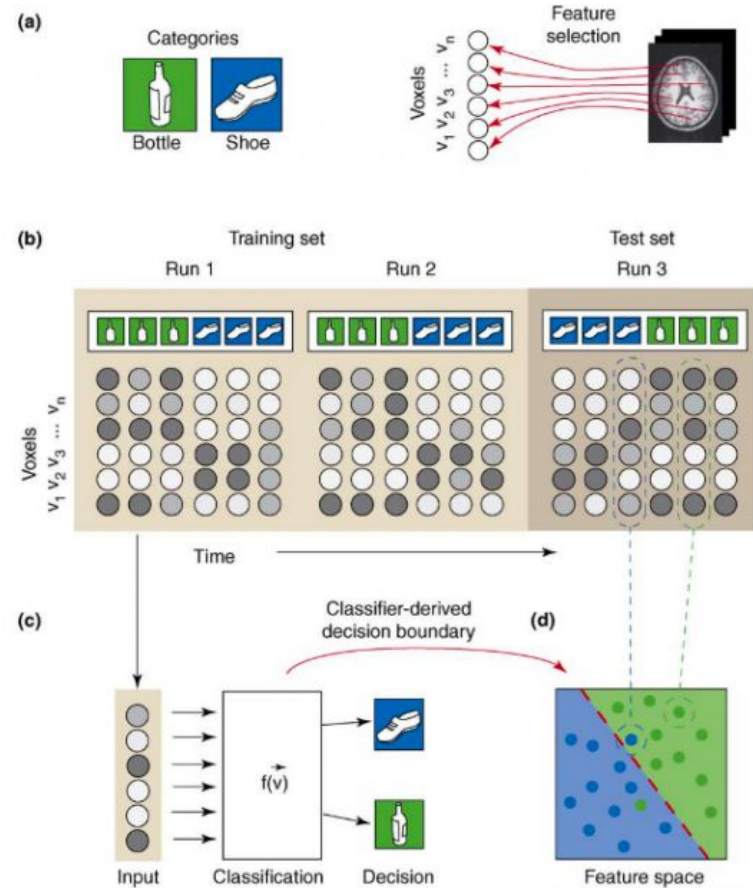


=SHOE?

During testing, the program must guess the object viewed on the basis of what it has learned about similar patterns of activity.

Multi-Voxel Pattern Analysis

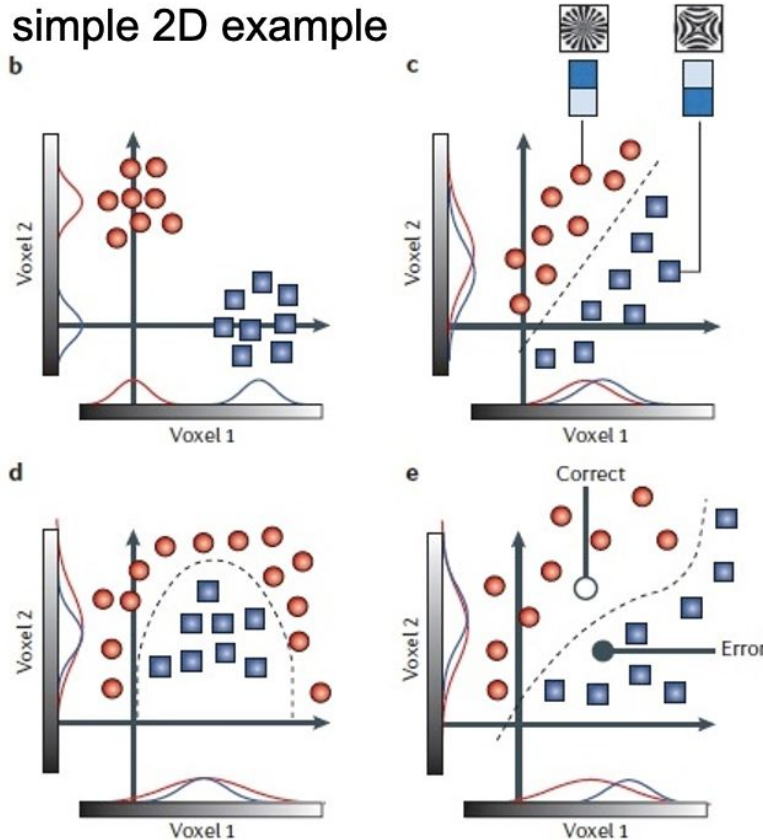
Supervised classification method where a classifier attempts to capture the relationships between **spatial pattern of fMRI activity** and **experimental conditions**.



simple 2D example

Classifier can act on single voxels.
Conventional fMRI analysis would detect the difference.

Classifier would require curved decision boundary

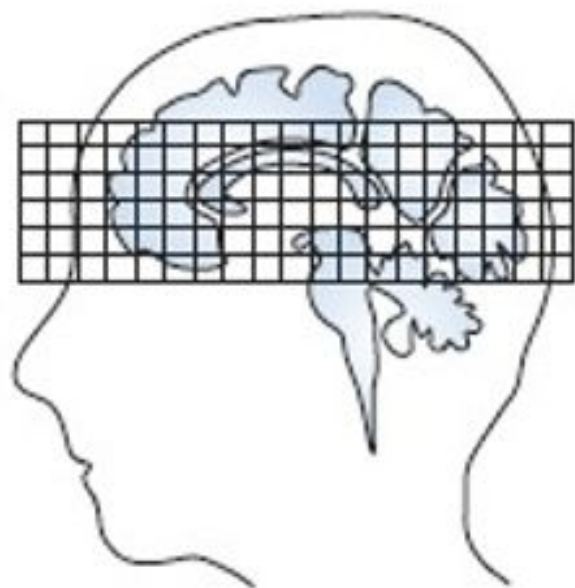


Classifier can not act on single voxels because distributions overlap

Classifier can act on combination of voxels using a linear decision boundary

White and black circles show examples of correct and erroneous classification in the test set

a 9 voxels \rightarrow 9 dimensions



N = 9

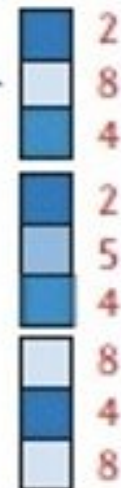
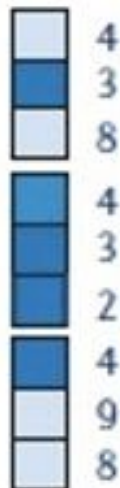


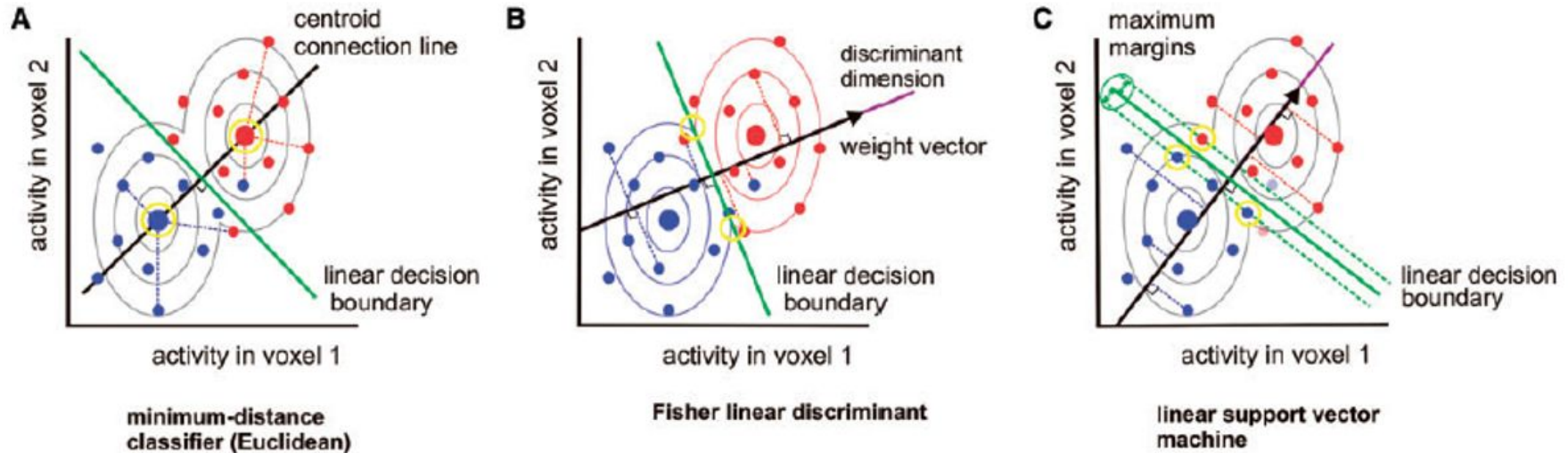
Image 1



Image 2



Linear support vector machine (SVM) is the most common approach to decide how to draw the line.

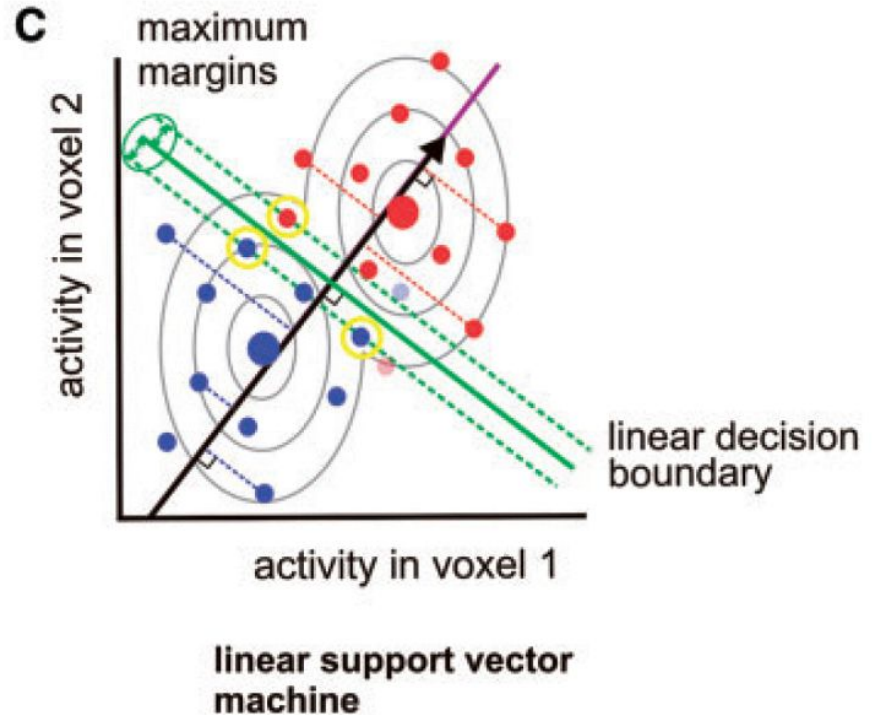


Support Vector Machine (SVM)

SVM finds a linear decision boundary that discriminates between two sets of points

Constrained to have the largest possible distance from the closest points on both sides.

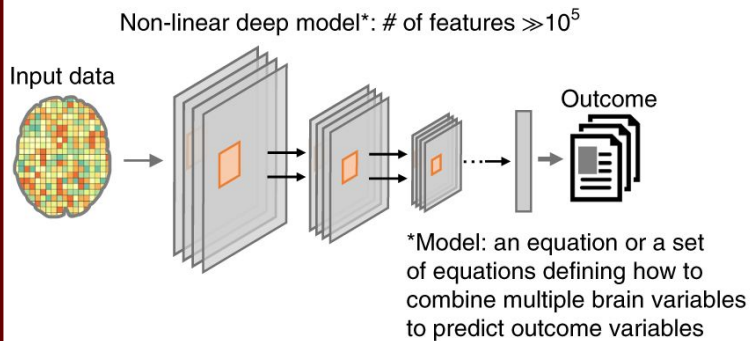
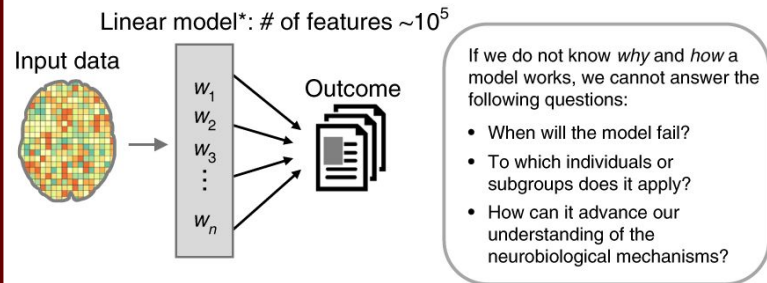
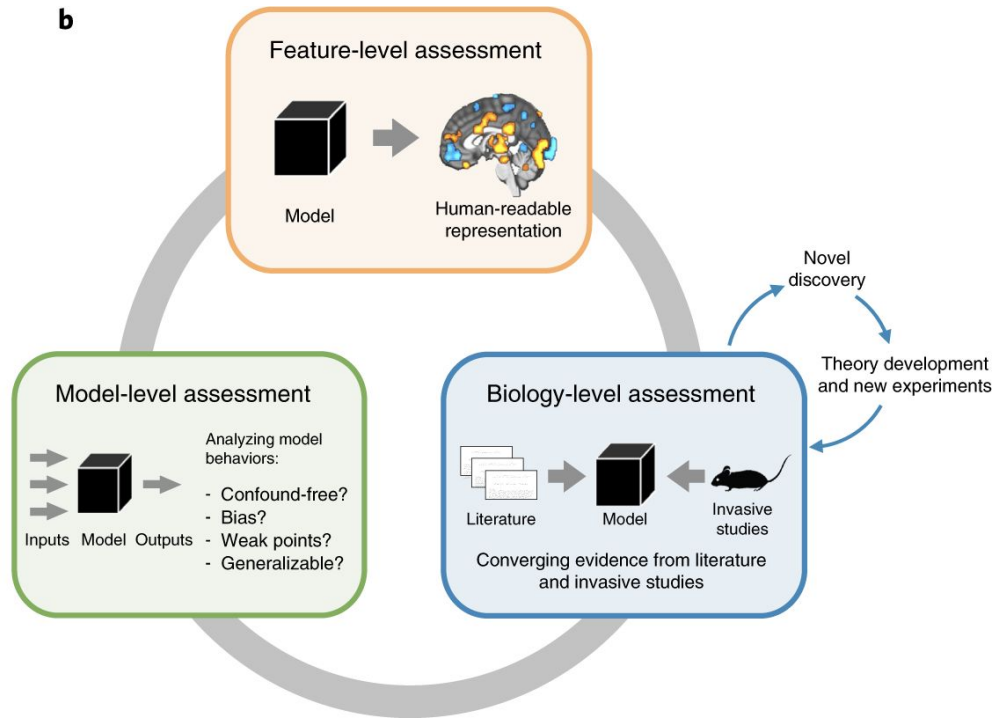
Response patterns closest to the decision boundary (yellow circles) that defined the margins are called “support vectors”.

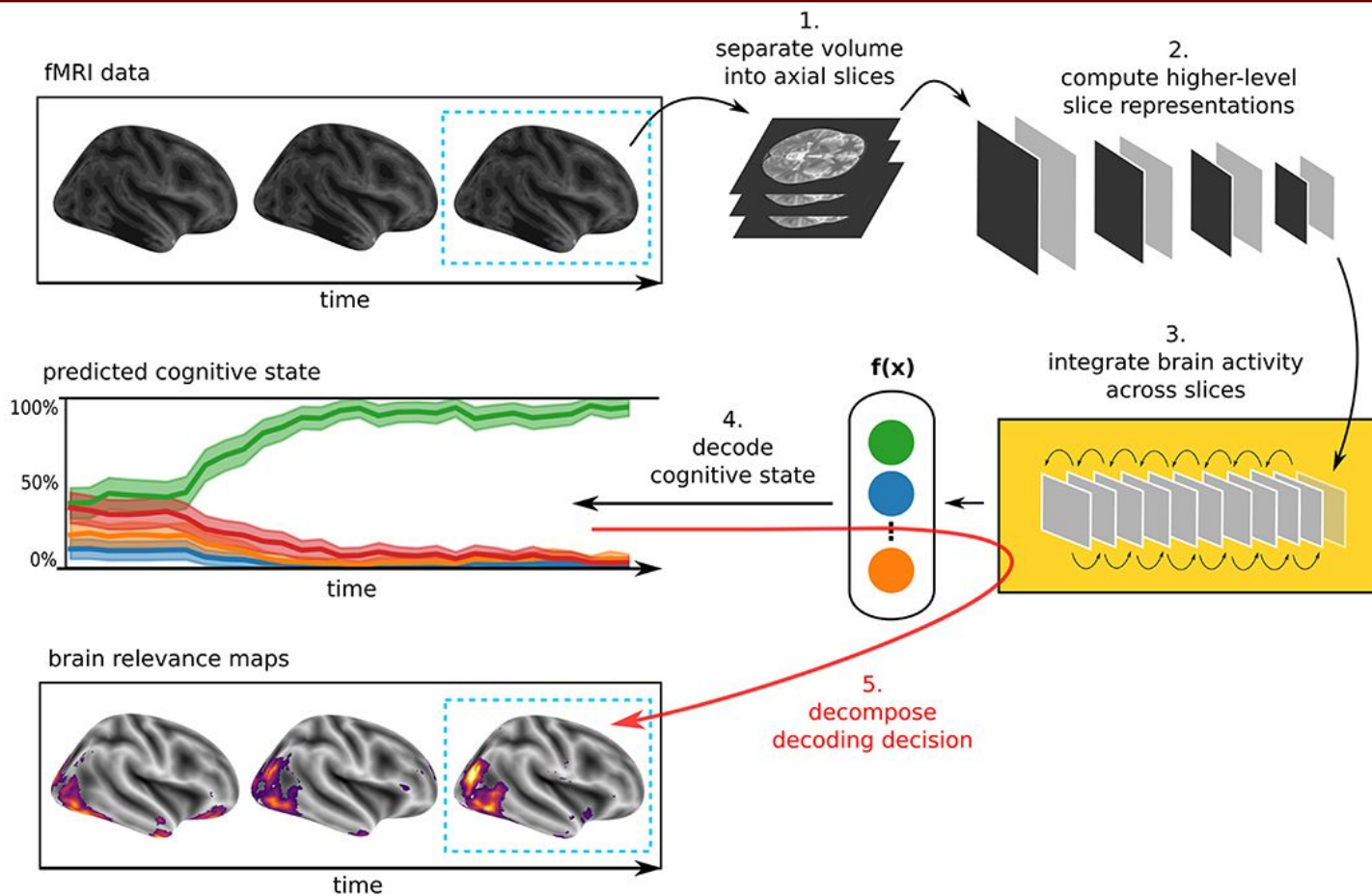


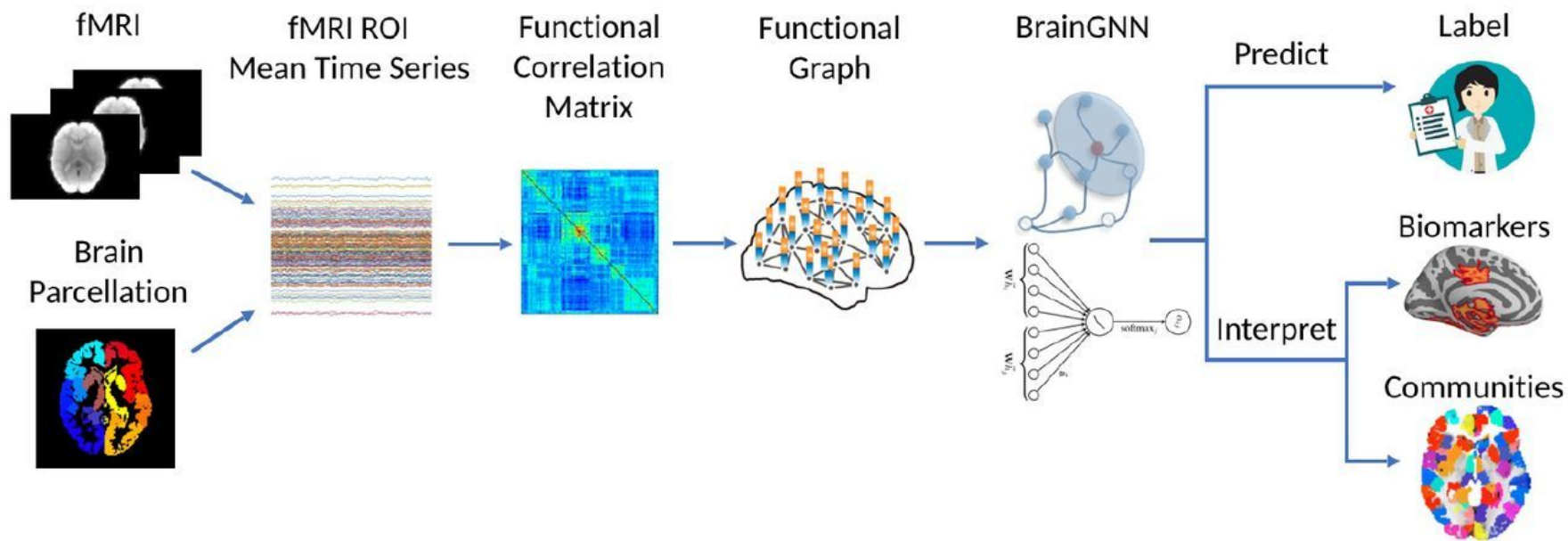
Steps to use classifiers

1. Get an estimate of activation for each voxel in ROI
2. Break data into independent training and test sets
3. Train classifier on training set, test on test set to determine accuracy of classification for each subject
4. Test significance of classification

Advanced Algorithms

a**b**





Appendix

fMRI Experimental Design

