

# **Pattern-information fMRI**

**stimulus decoding, computational-model testing,  
columnar sensitivity**

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# Overview

## Part A: Pattern-information fMRI

- **Neuroscientific motivation**
- **Insights on object vision**

## Part B: Recent studies from my lab

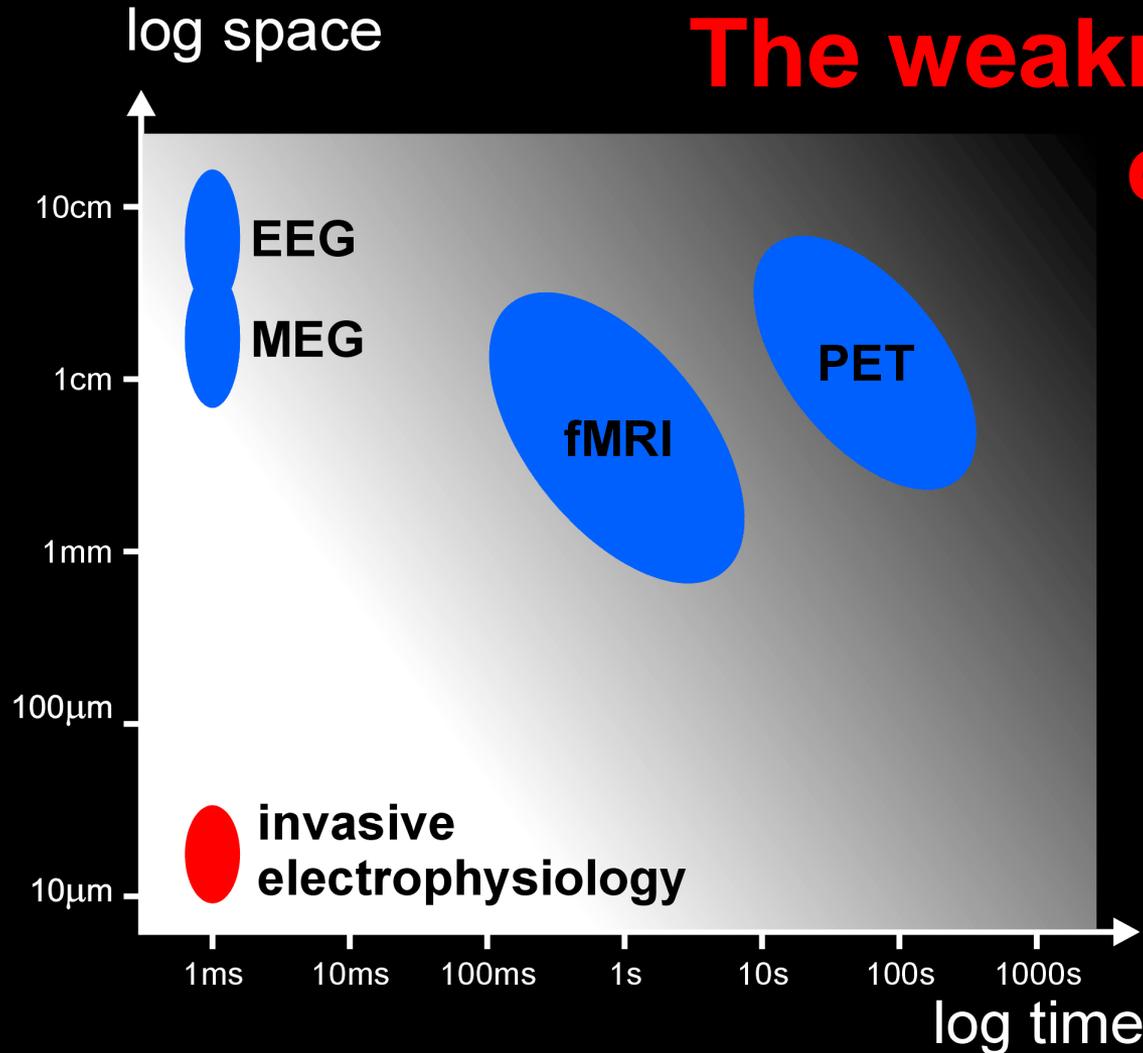
- **Columnar sensitivity?**  
**Origins of V1 orientation signals in fMRI**
- **Individually unique object representations**

# Part A: Learning objectives

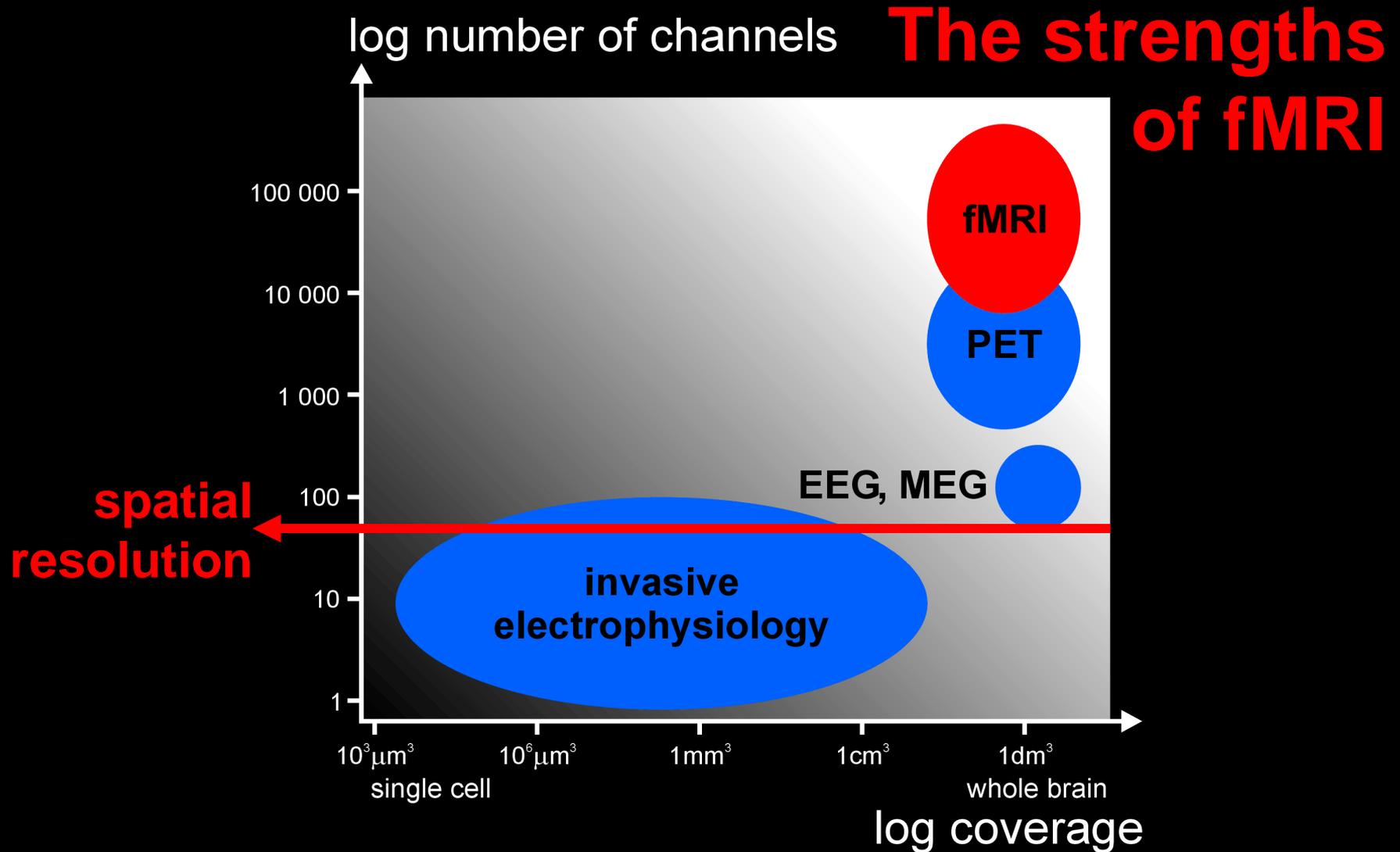
- 1 Why fMRI?**
- 2 Why pattern-information fMRI?**
- 3 Why decode fMRI?**
- 4 Why test encoding models with fMRI?**
- 5 Why should your analysis be at once data and hypothesis-driven?**

# 1 Why fMRI?

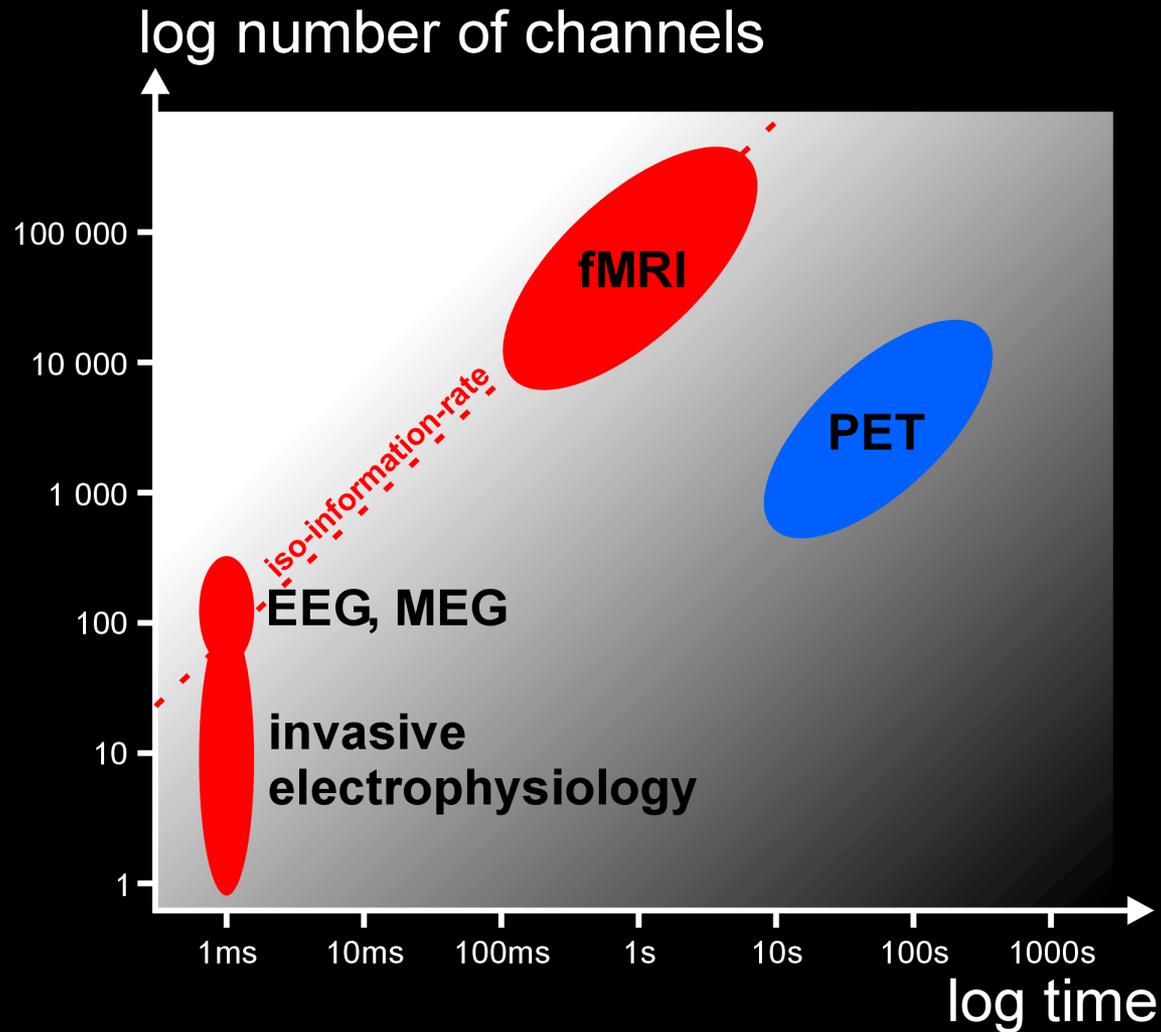
# Spatial and temporal resolution



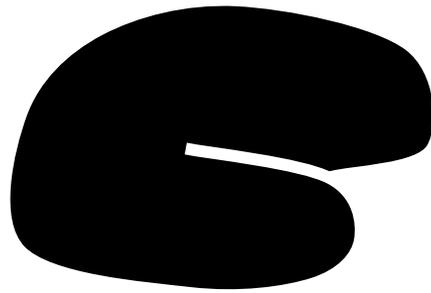
# Number of channels and coverage



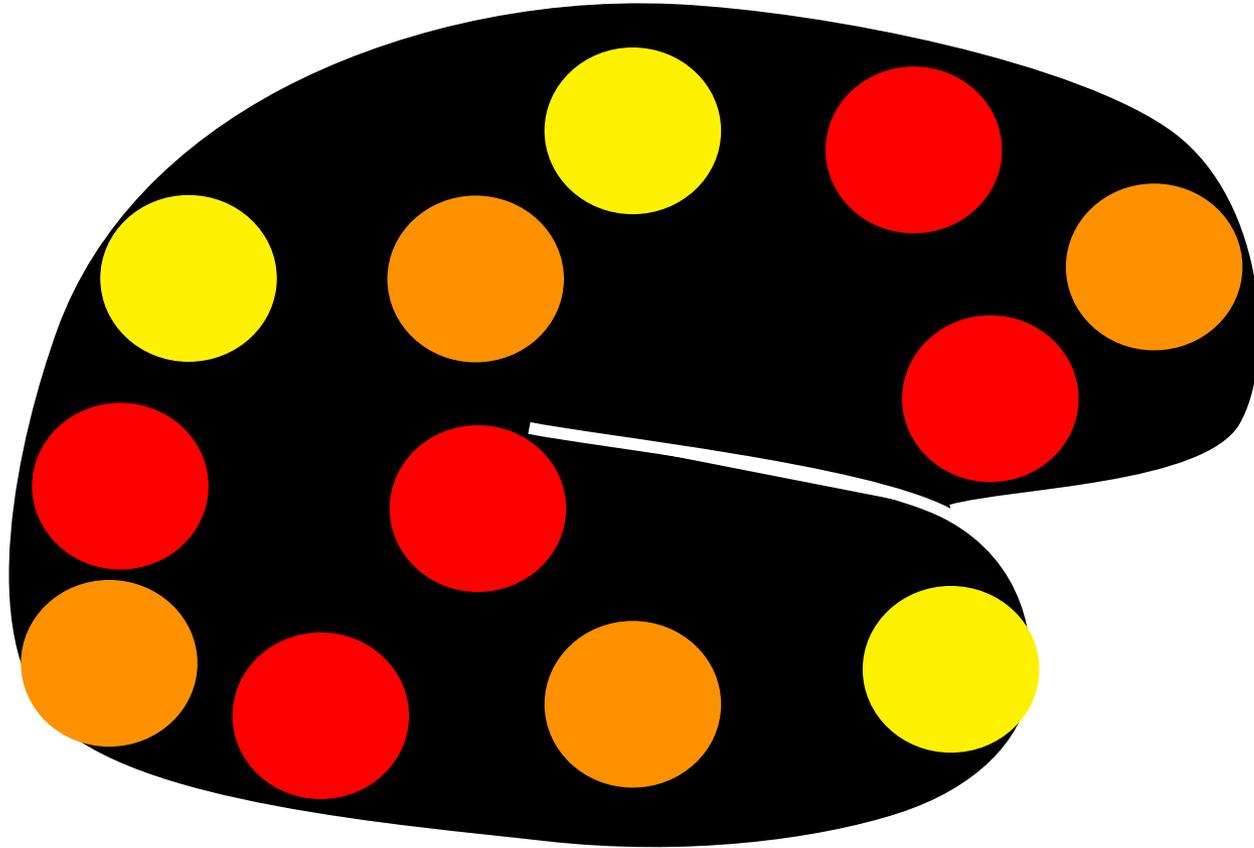
# Similar information rates



## **2 Why pattern-information fMRI?**

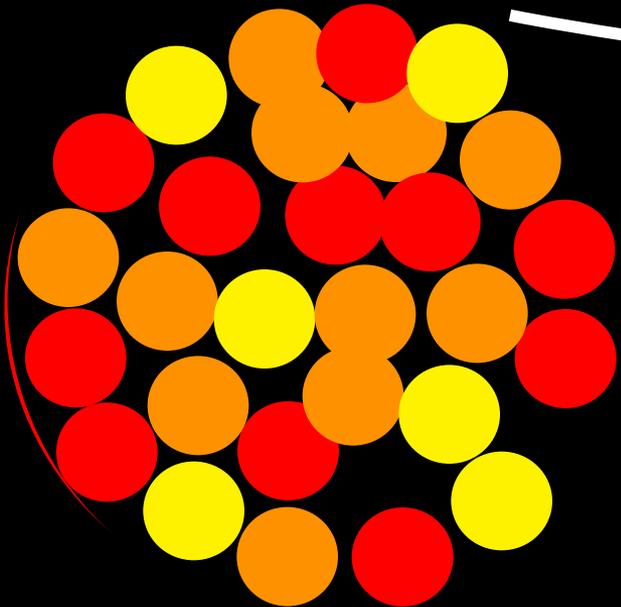


# Activation analysis



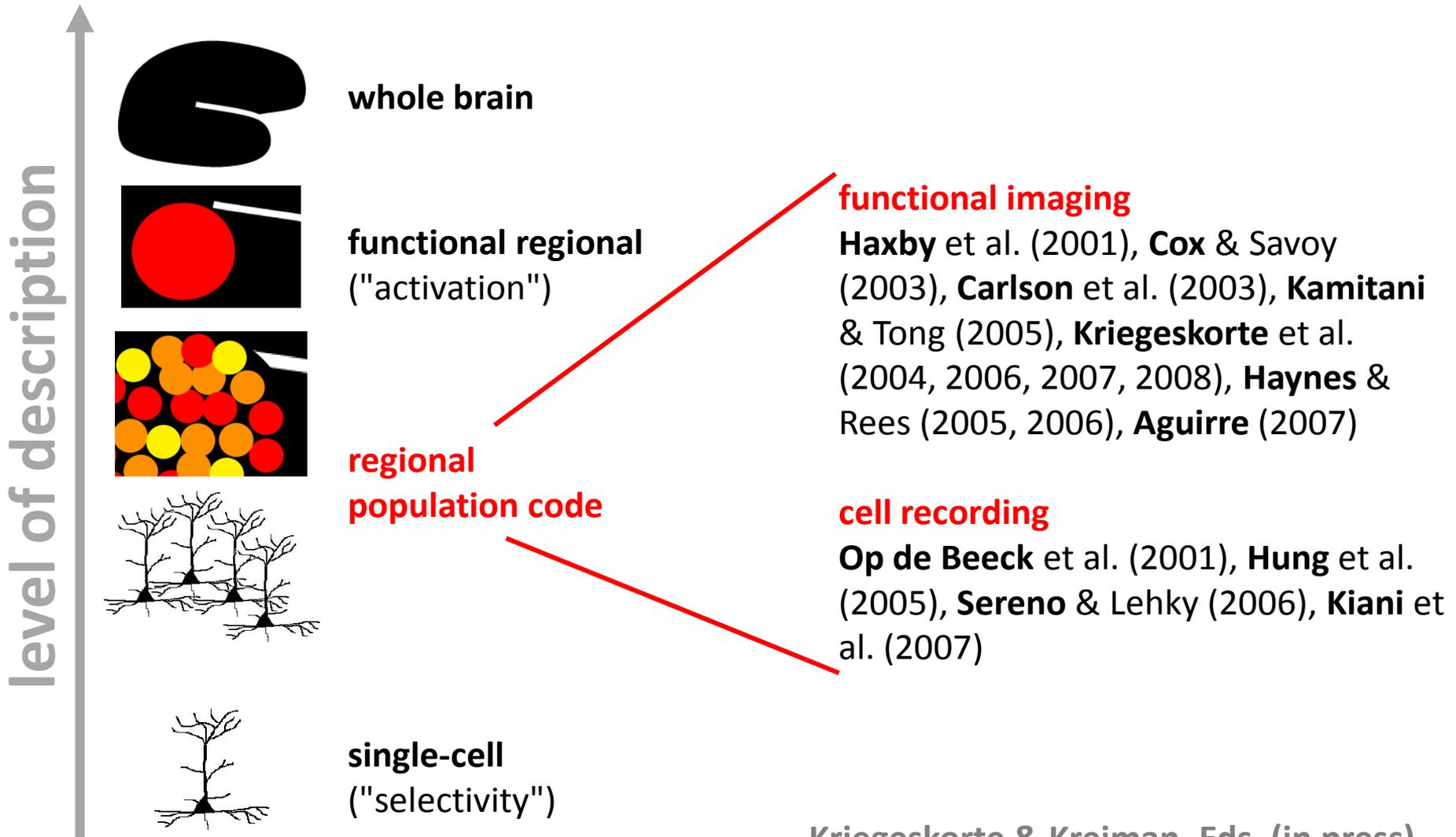
activation → "involvement" in cognitive function

# Pattern-information analysis



information →  
"representational content"

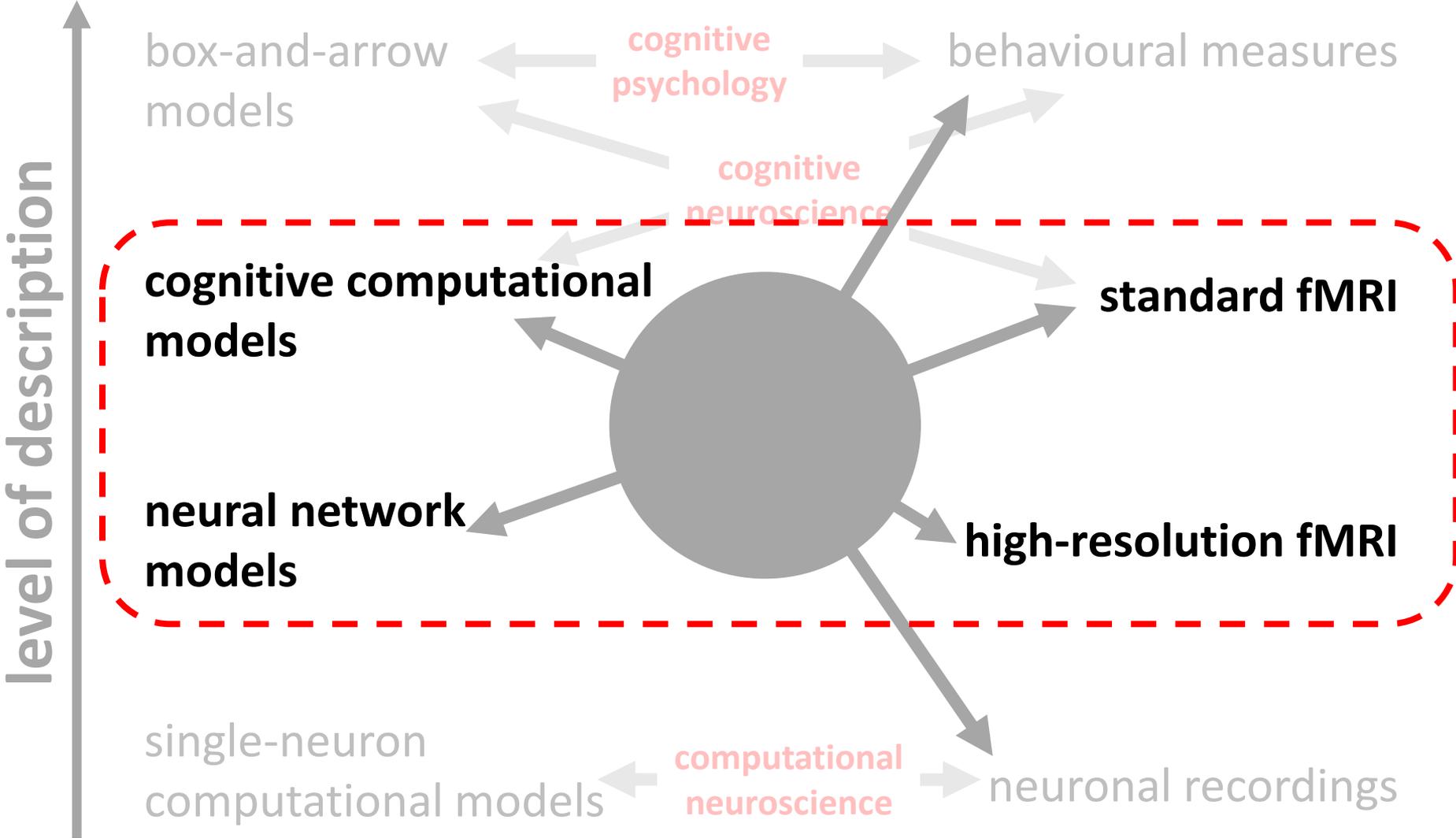
# Filling the explanatory gap





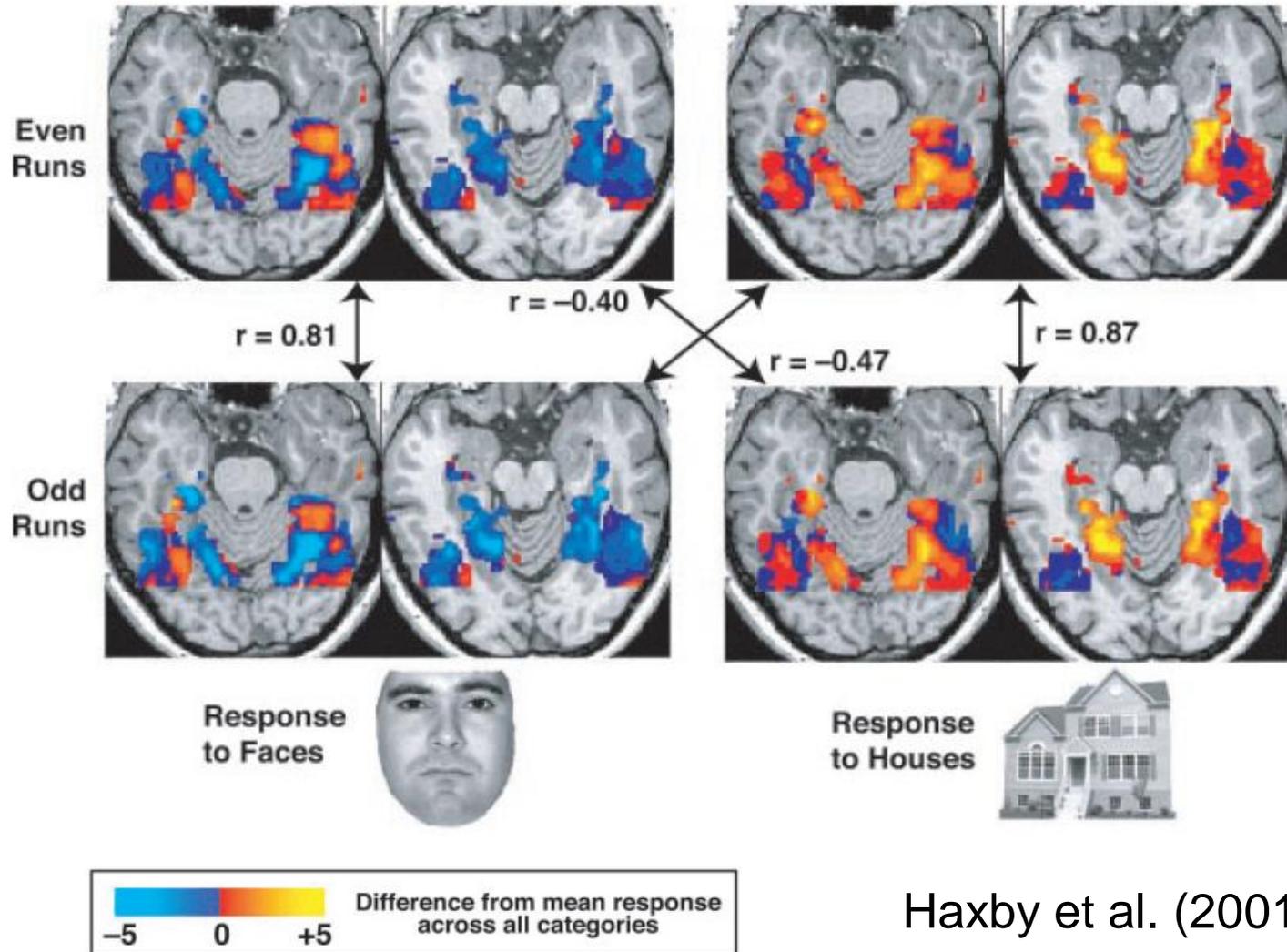
# theory

# experiment



# **3 Why decode fMRI?**

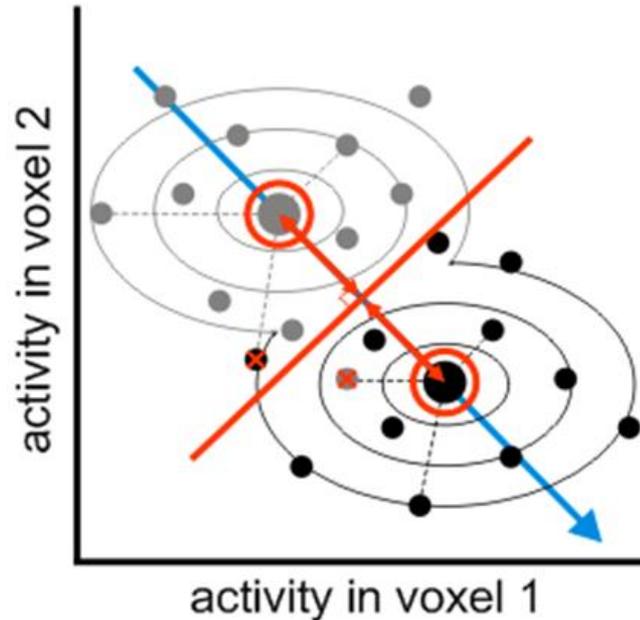
# “Decoding” stimuli from brain responses



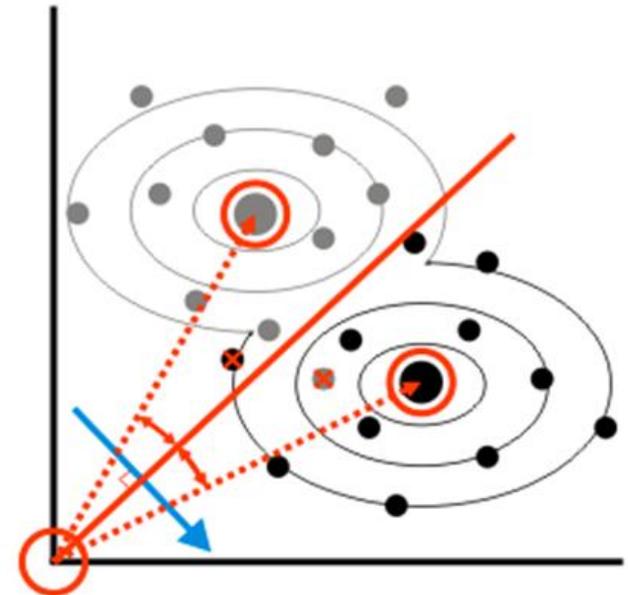
Haxby et al. (2001)

# Linear pattern-classifier analysis

**a** minimum-distance classifier

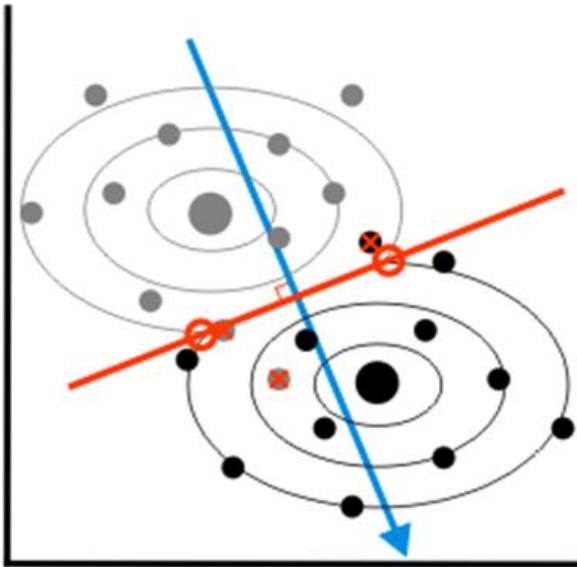


**b** maximum-correlation classifier

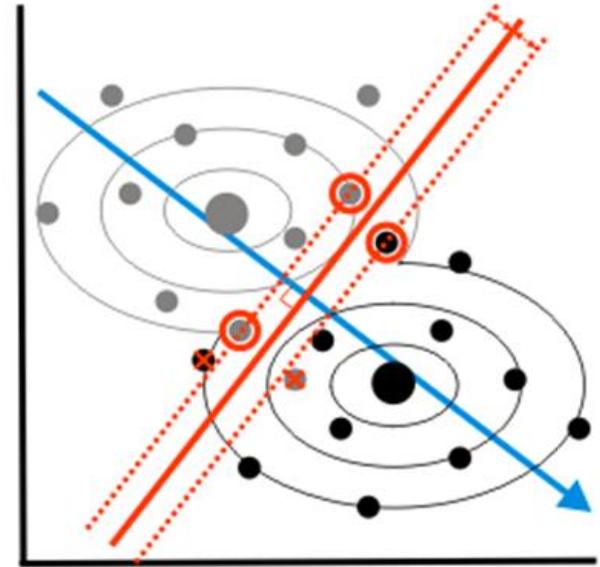


# Linear pattern-classifier analysis

**c** Fisher linear discriminant



**d** linear support vector machine



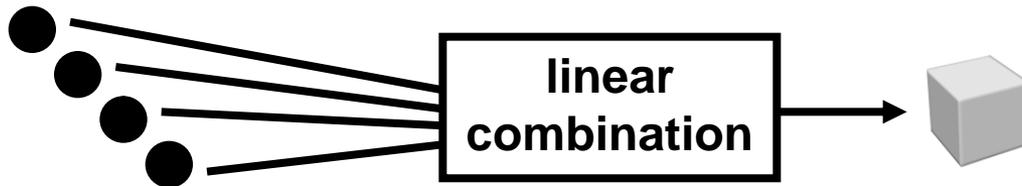
# Nonlinear classifiers?

- Nonlinear decoding typically doesn't generalise as well in fMRI (Misaki et al. 2010).
- Linearly decodable information can be considered “explicit” in the code.

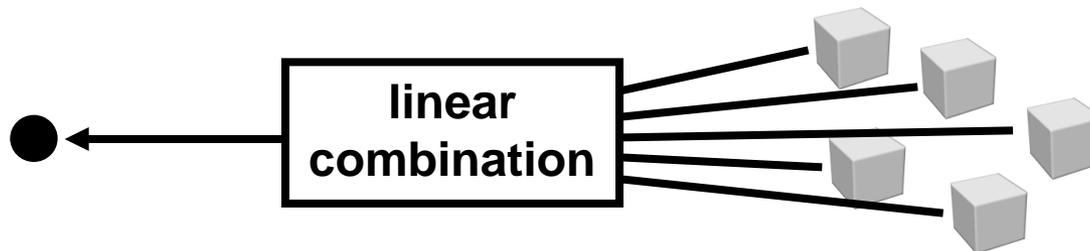
# Decoding or encoding?

stimulus

response



univariate  
activation  
analysis



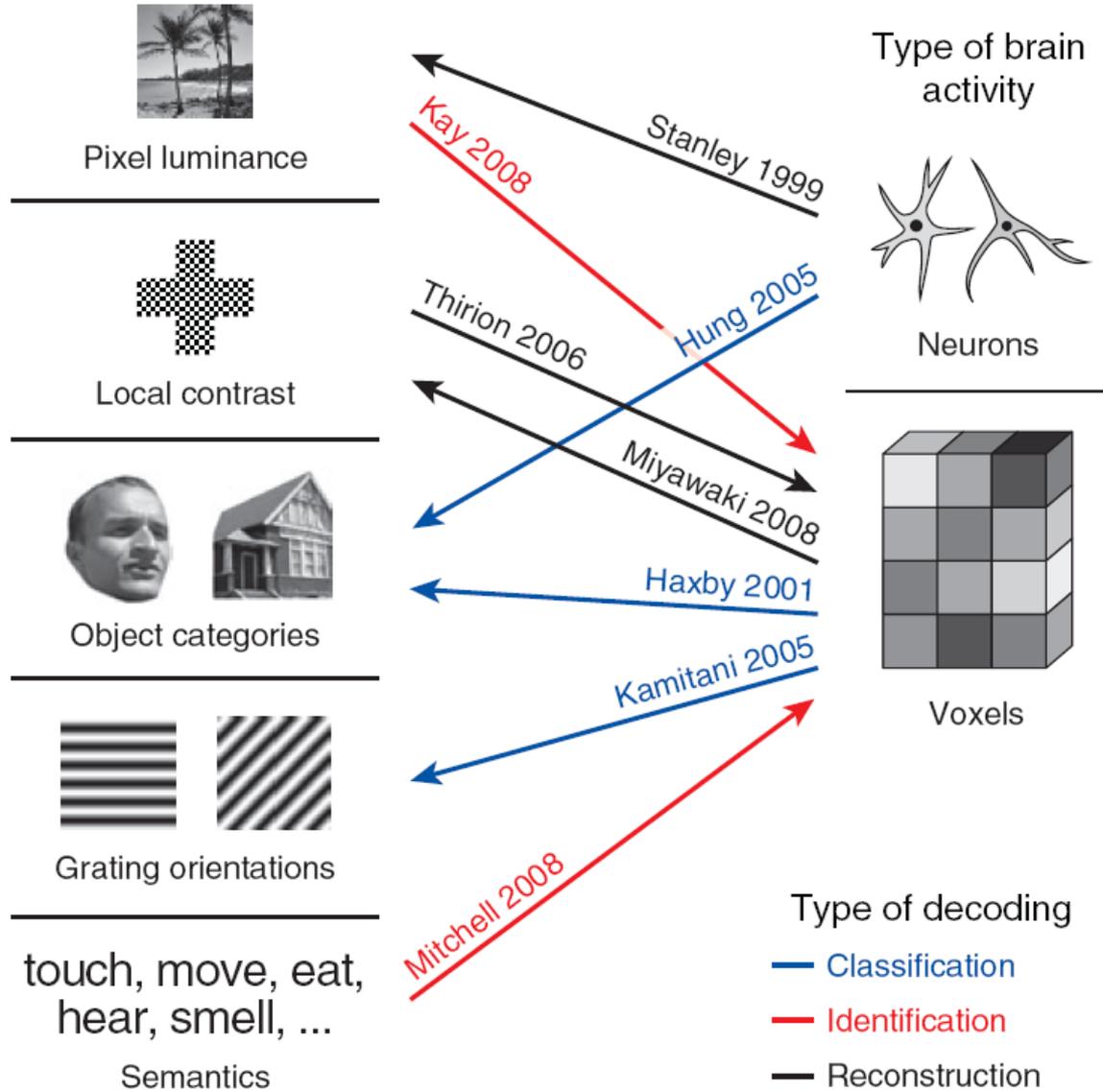
multivariate  
pattern-information  
analysis

"decoding", "prediction"

← spin

↑  
substance

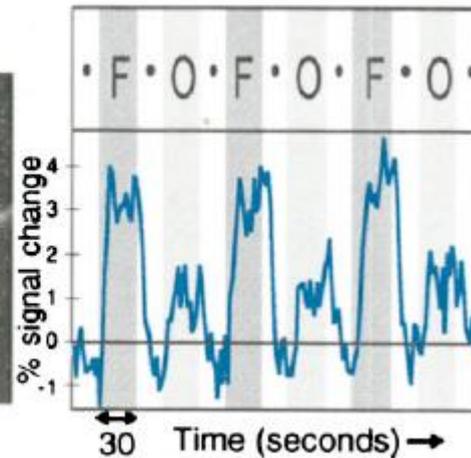
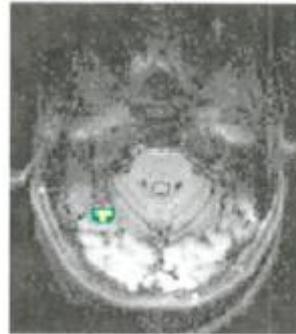
# Stimulus representation



from Kay & Gallant (2009)

# “Predicting” brain responses from stimuli

4a. Faces > Objects



Kanwisher et al. (1997)

# Potential title claims

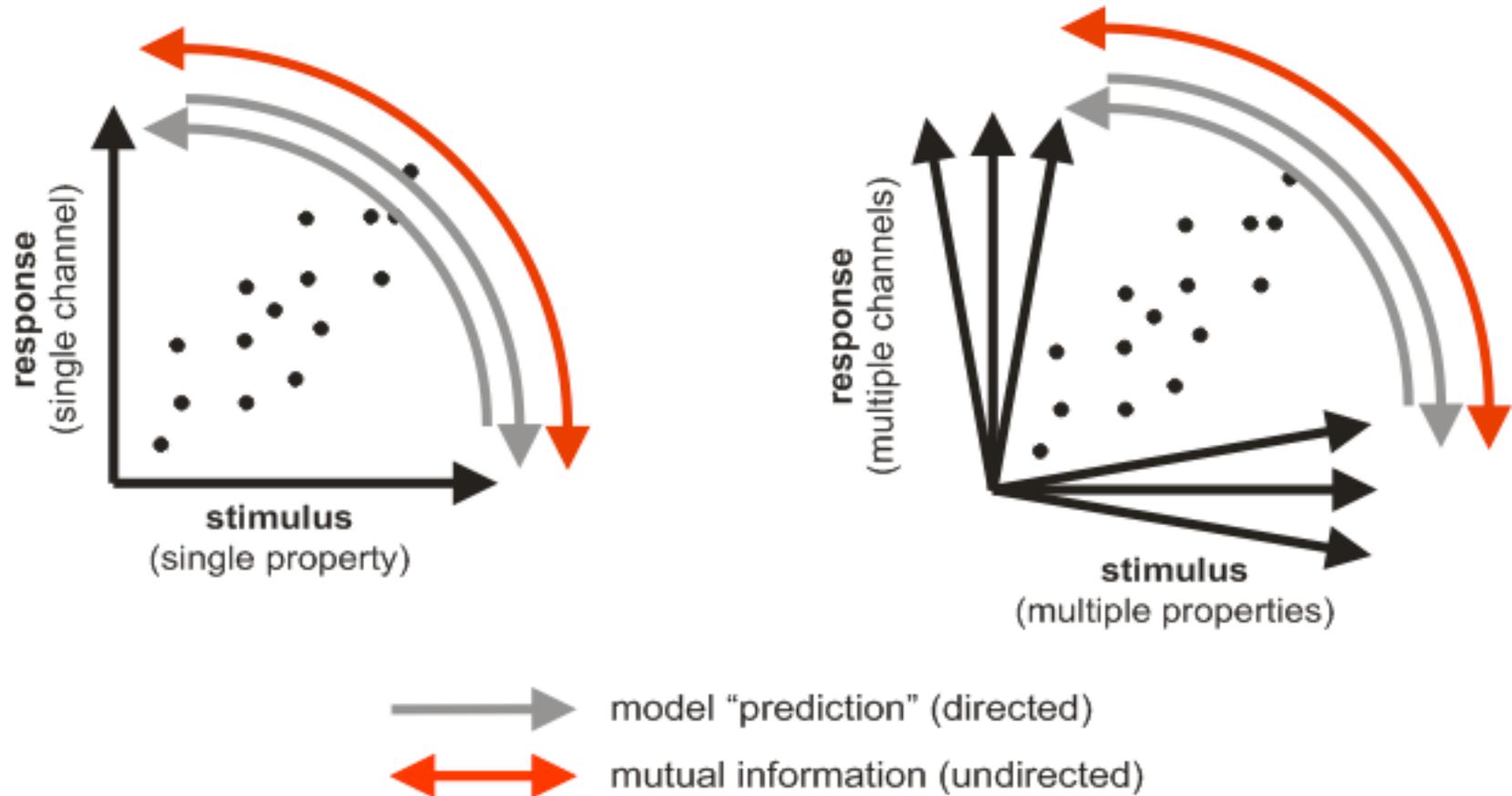
Fusiform region responds more strongly to visual faces than objects. *Neuroreport*, 1997.

**The fusiform face area: a module in human extrastriate cortex specialized for face perception. *J Neurosci*, 1997.**

Brain reading: Predicting face percepts from fusiform activity. *Science*, 1997.

**Model direction has no neuroscientific  
implications**

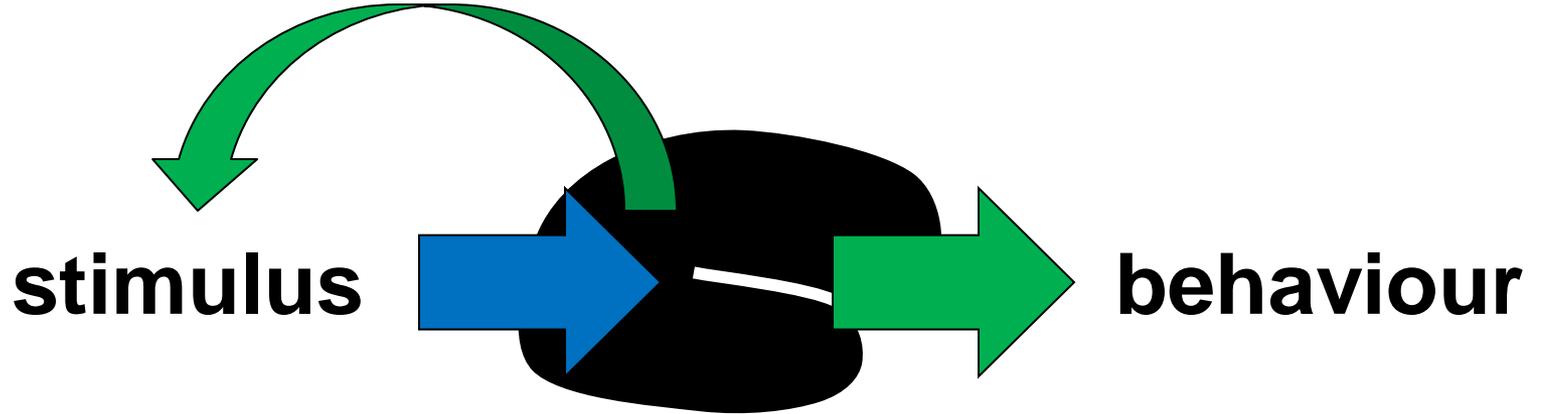
# Model direction has no neuroscientific implications



## **4 Why test encoding models with fMRI?**

**decoding**

**encoding**

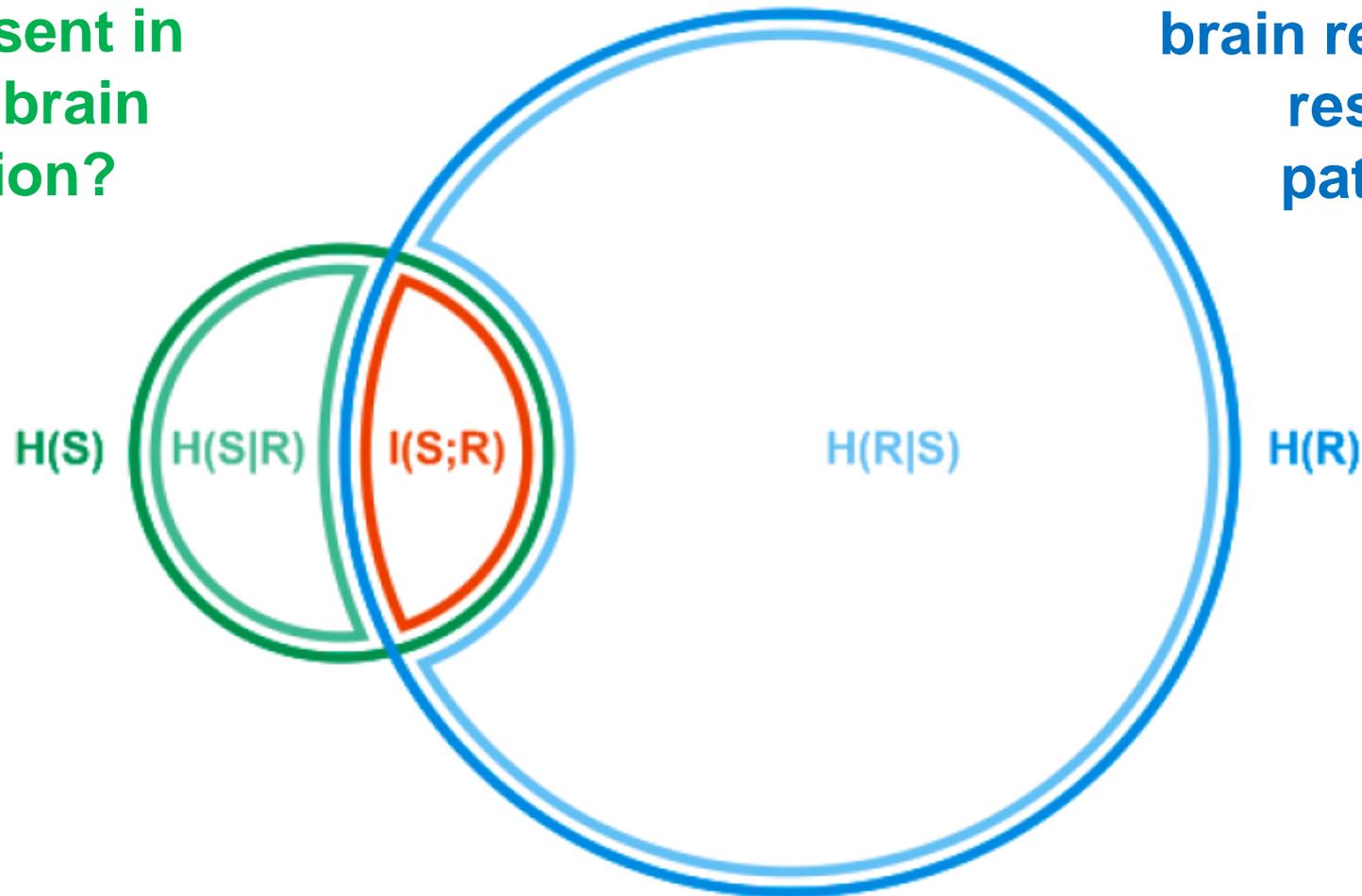


# decoding

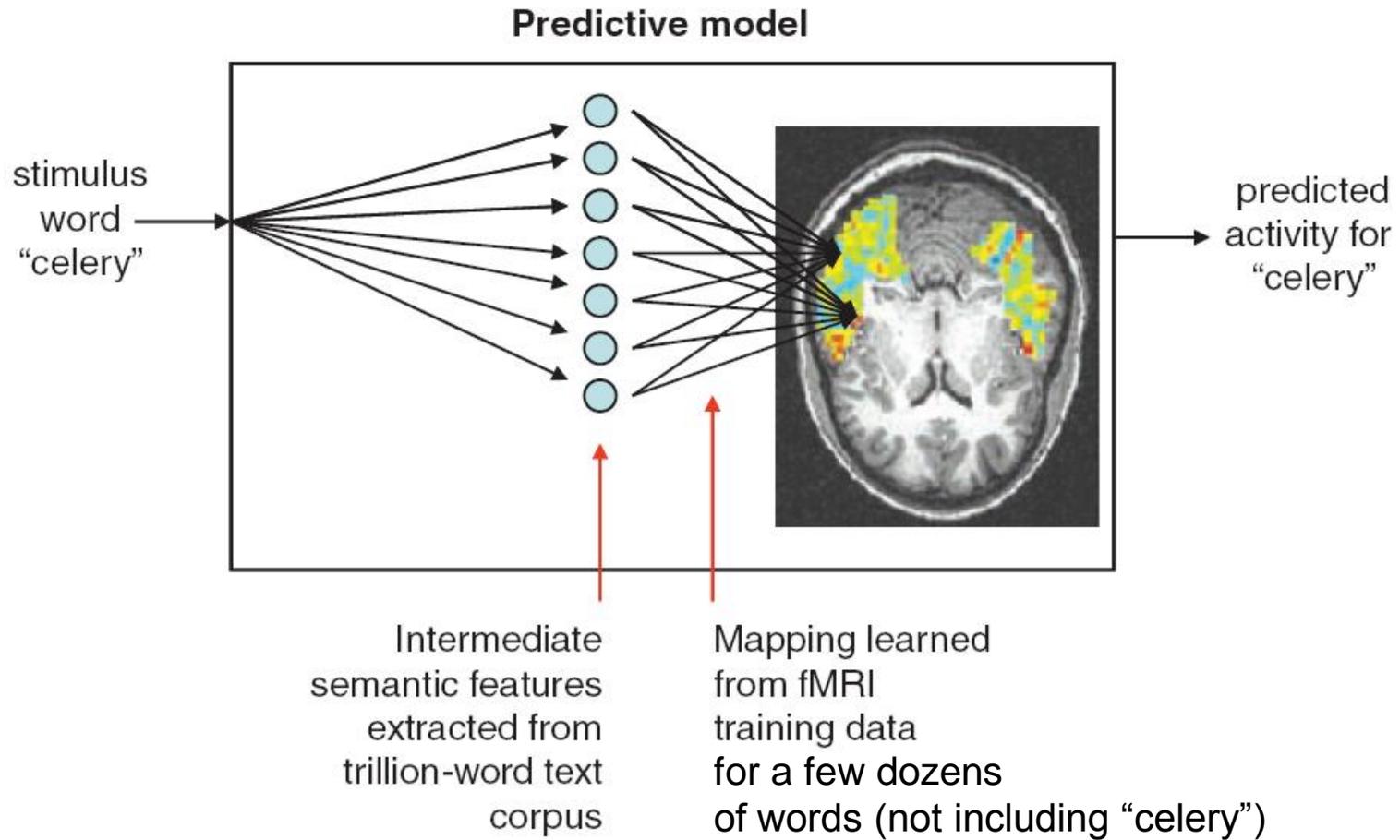
How much stimulus information is present in the brain region?

# encoding

How well does our model explain the brain region's response patterns?

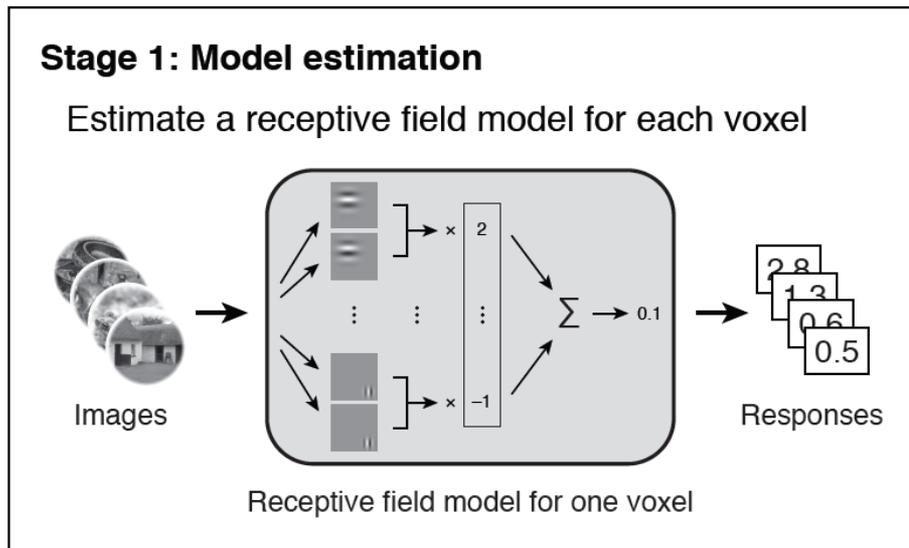


# Predicting brain responses for *novel* stimuli



Mitchell et al. (2008)

# “Predicting” *novel* stimuli from brain responses

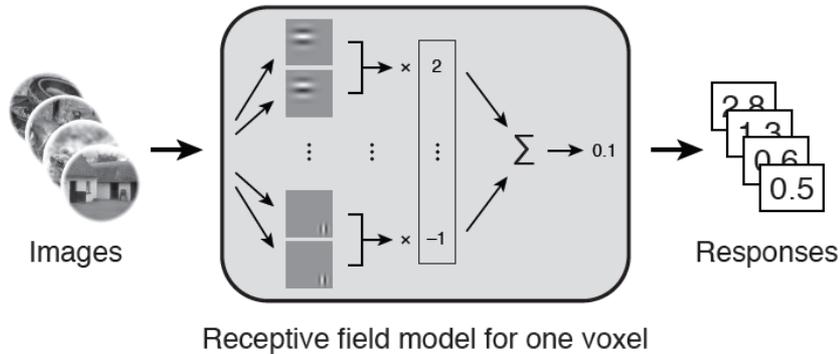


Kay et al. (2008)

# “Predicting” *novel* stimuli from brain responses

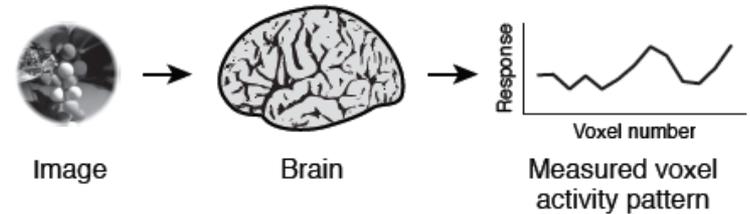
## Stage 1: Model estimation

Estimate a receptive field model for each voxel

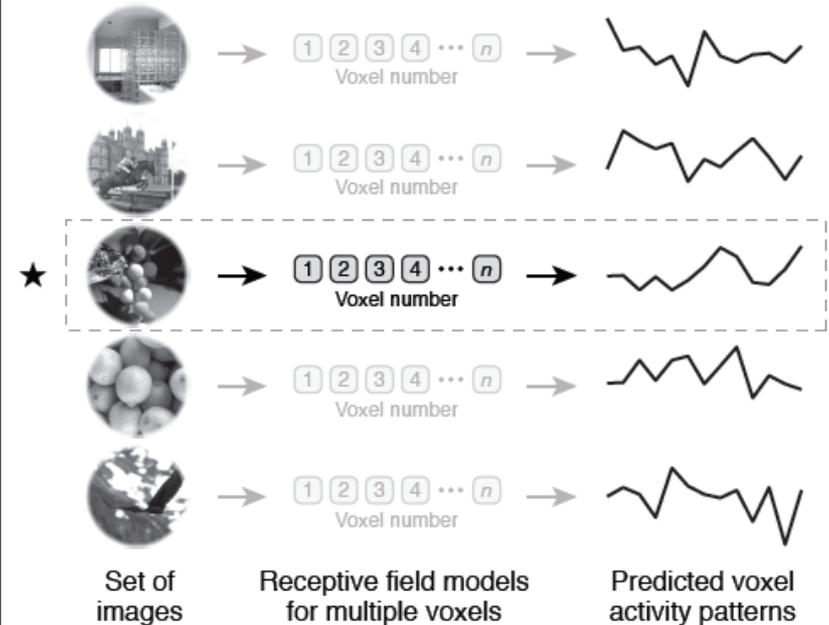


## Stage 2: Image identification

(1) Measure brain activity for a *novel* image



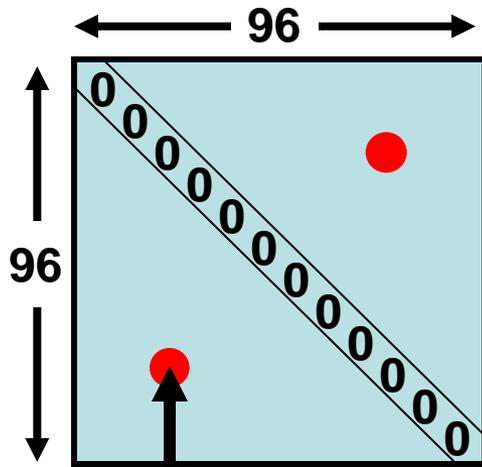
(2) Predict brain activity for a set of images using receptive field models



(3) Select the image (★) whose predicted brain activity is most similar to the measured brain activity

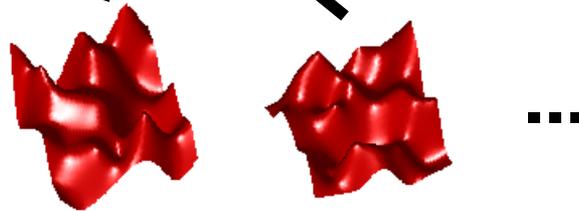
**Which distinctions between stimuli are reflected and which are lost in the population code?**

dissimilarity matrix



compute dissimilarity (1-correlation across space)

response patterns



stimuli

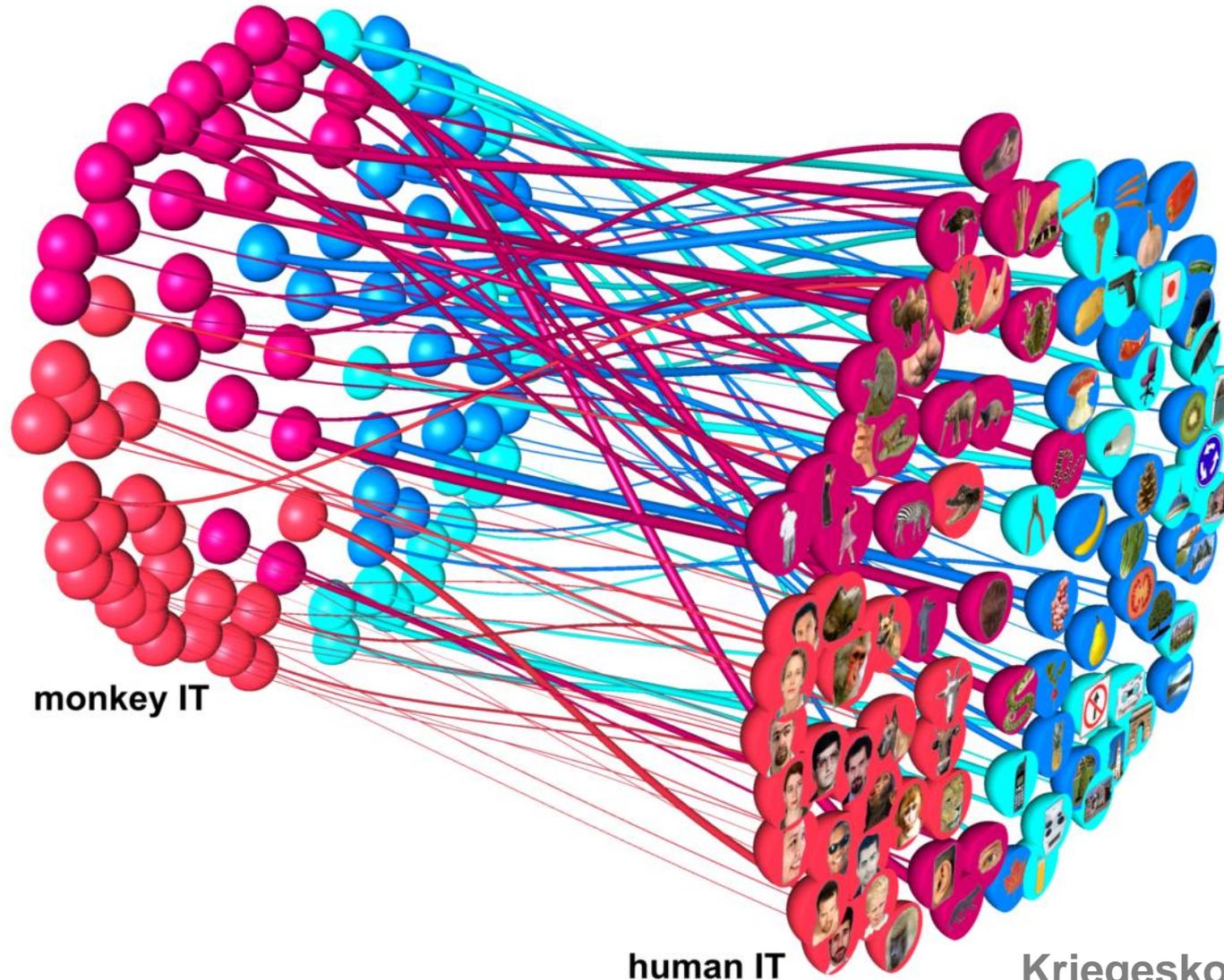


96

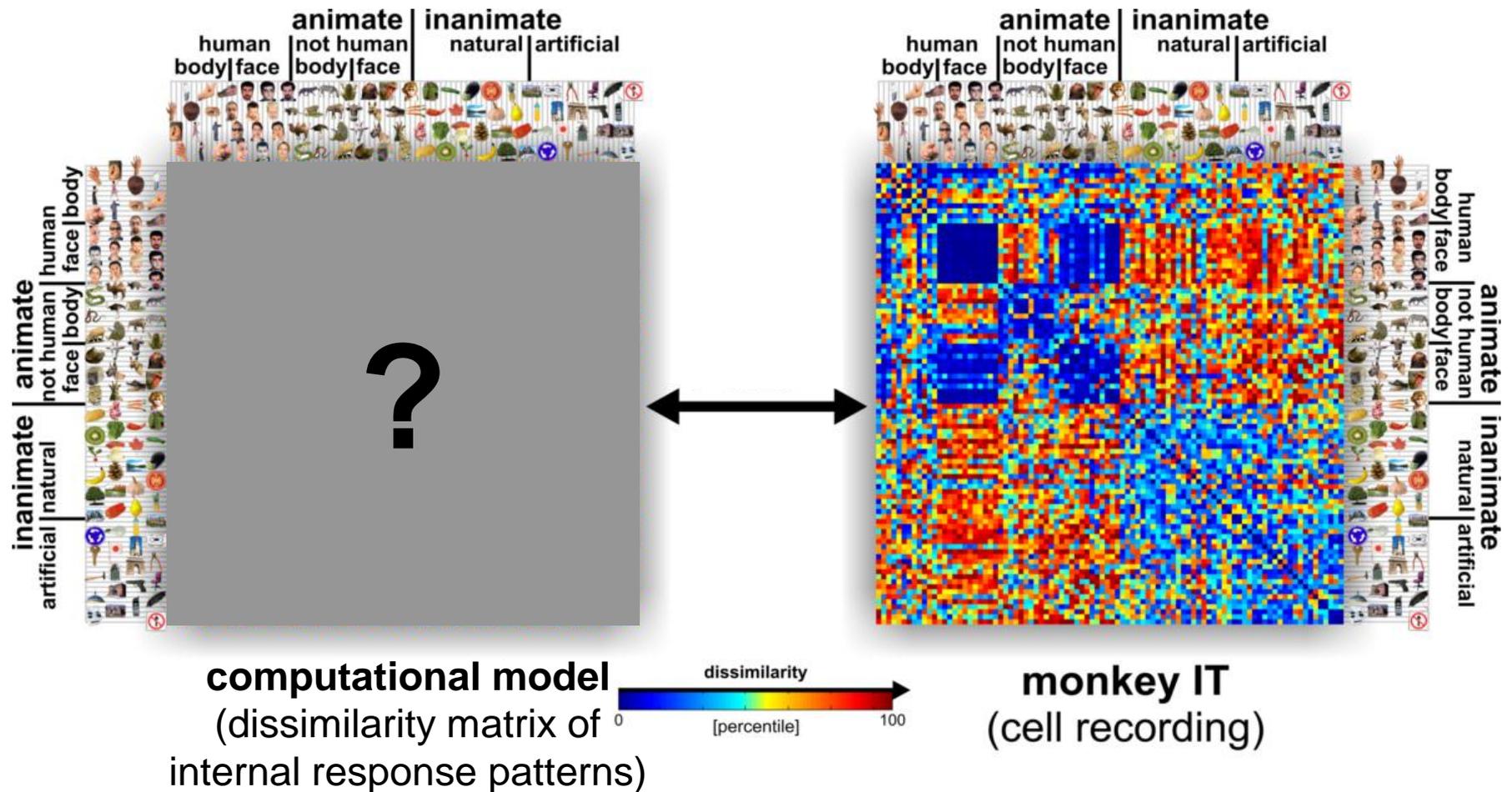


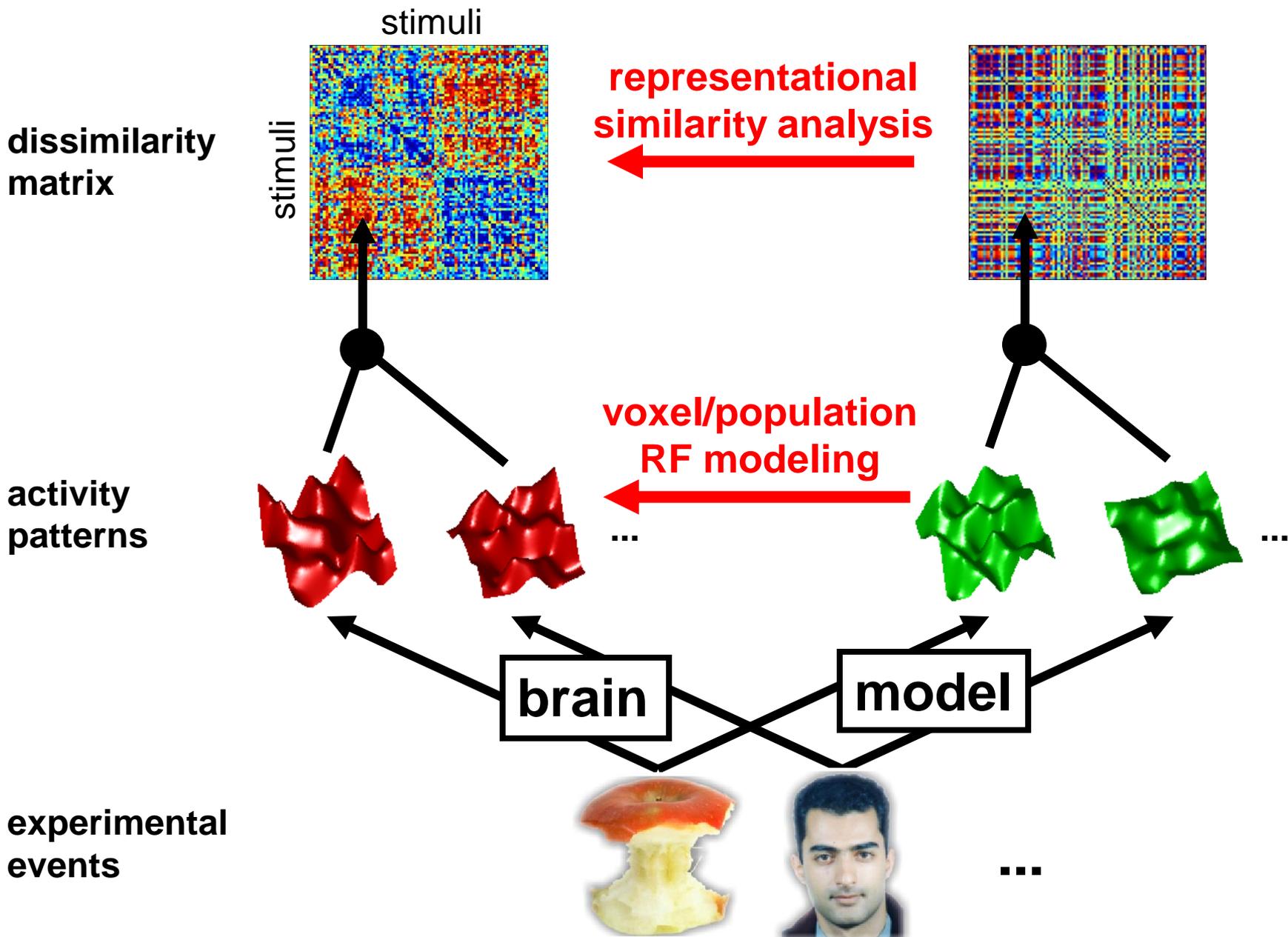


# Inferior temporal (IT) representational space



# Comparing brain and model representations

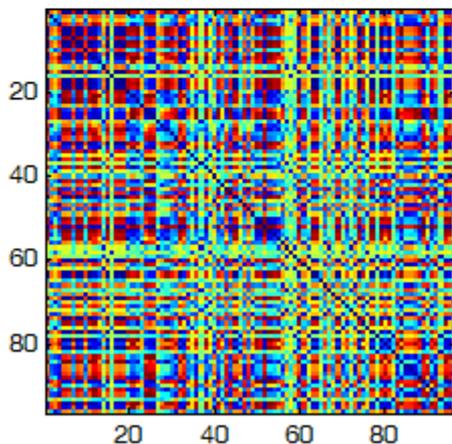




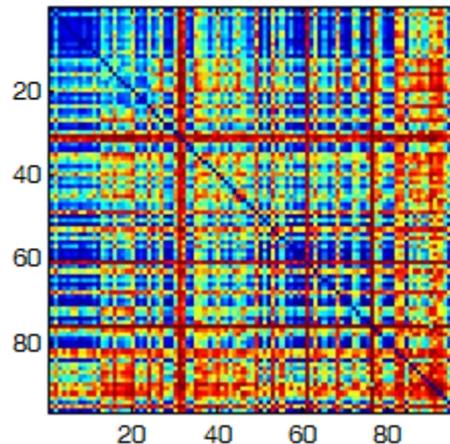
# Model dissimilarity matrices

# Model dissimilarity matrices

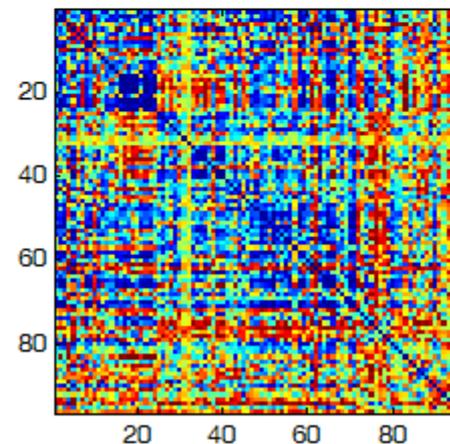
HMAX model



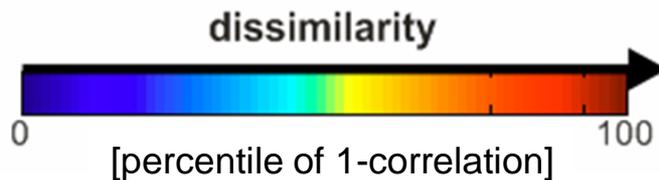
V1 model



stimulus image



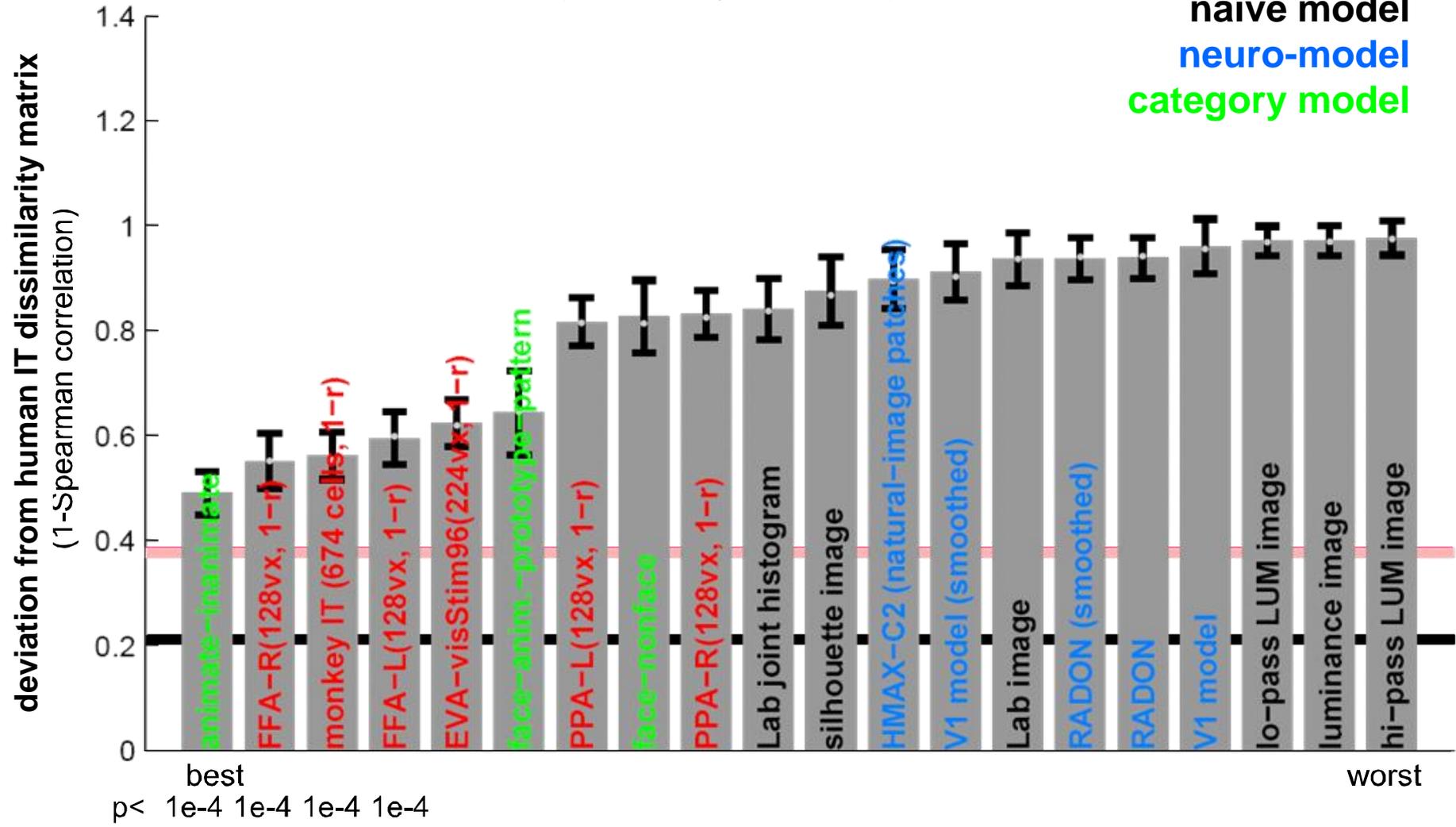
Riesenhuber & Poggio 2002

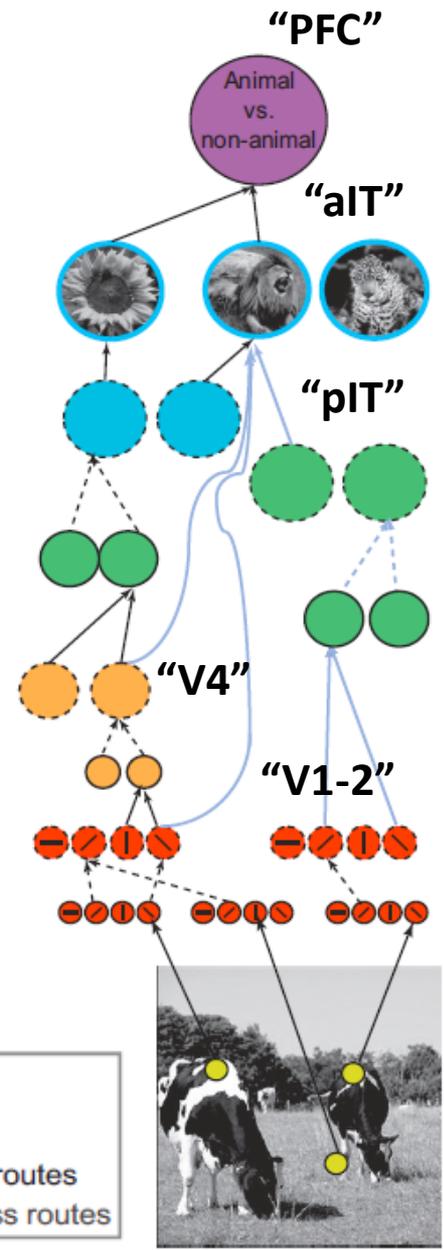
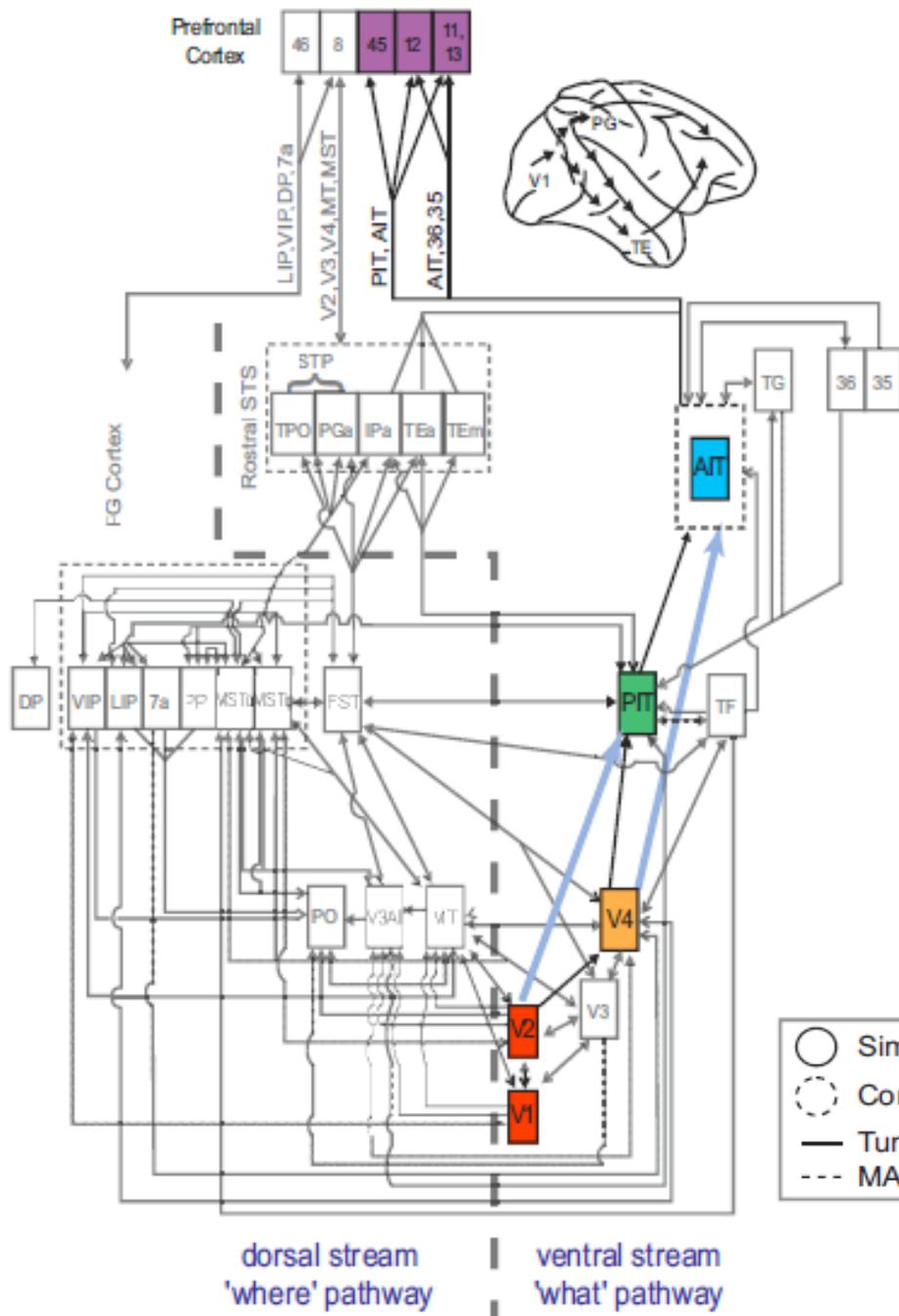


# Human IT

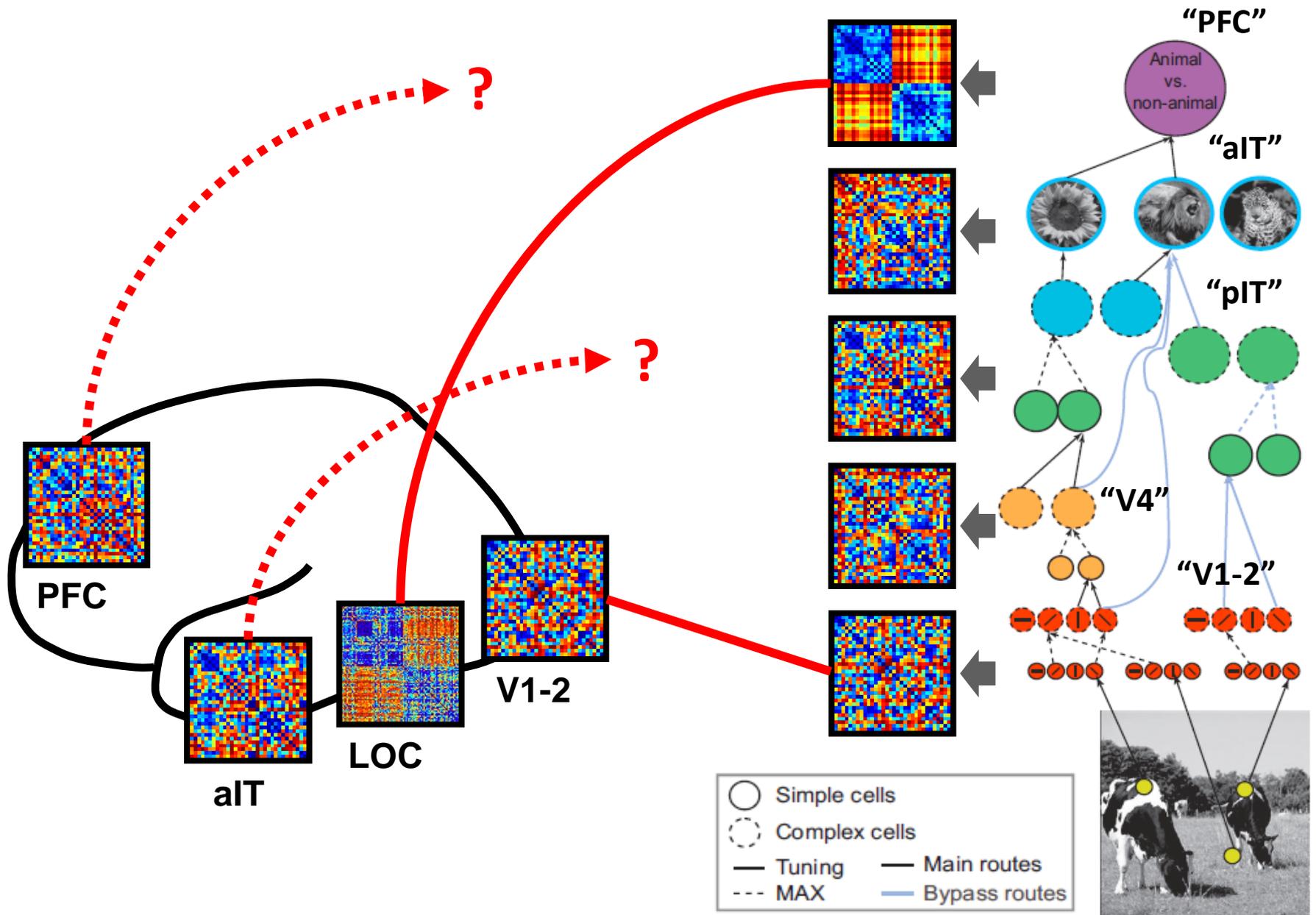
(dissimilarity SNR=1.29)

brain region  
naive model  
neuro-model  
category model





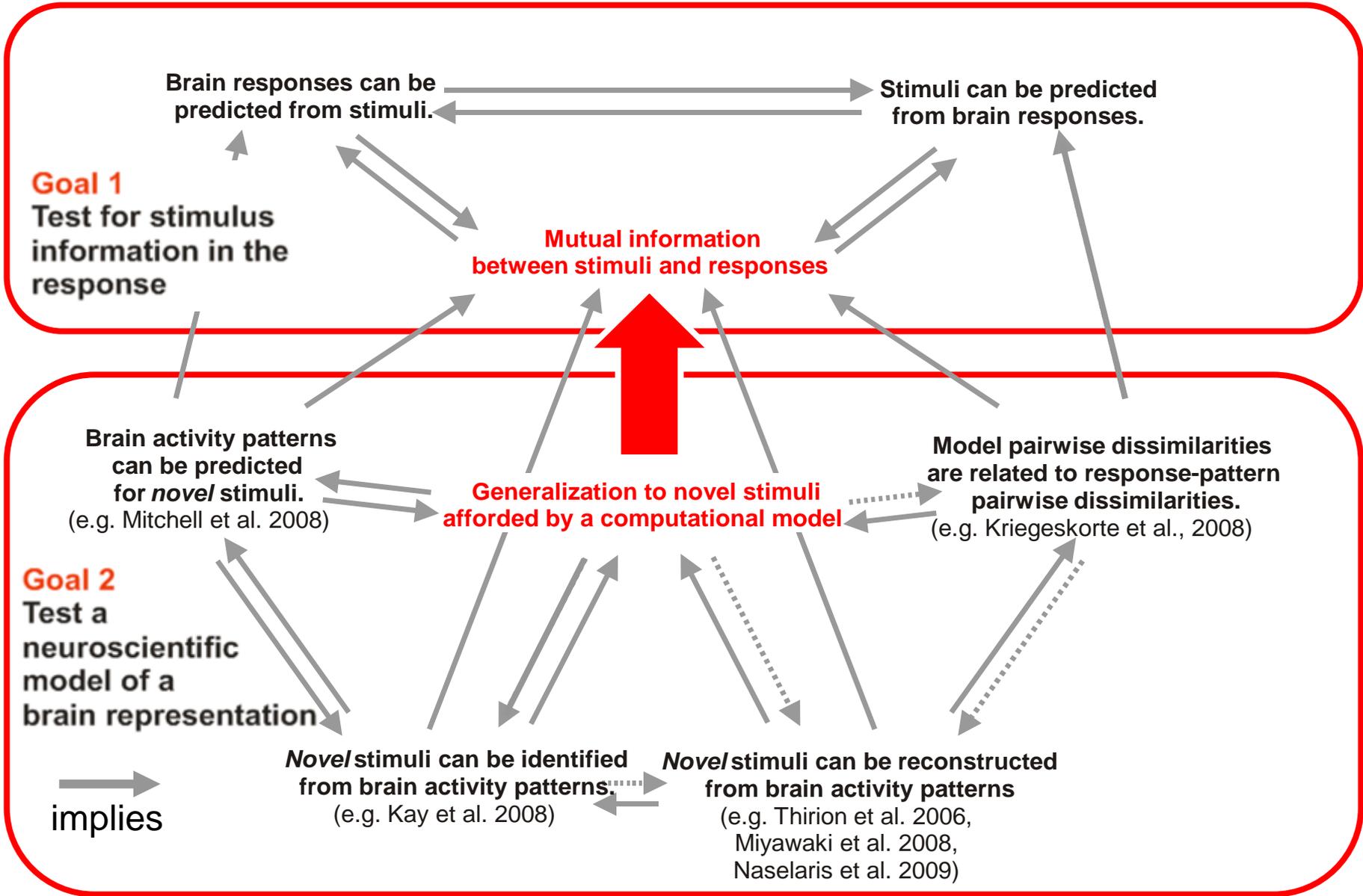
Serre, Oliva & Poggio (2007)  
PNAS



Serre, Oliva & Poggio (2007)  
PNAS



**What **neuroscientific implications**  
do these different types of  
pattern-information result have?**



**5 Why should your analysis be at once data- and hypothesis-driven?**



**broad sensitivity**  
(can find unexpected,  
strong effects)

**focussed sensitivity**  
(can find expected, weak  
effects distinguishing theories)

**data  
driven**

**hypothesis  
driven**

**fishing expedition**  
(not optimally sensitive  
to weak effects  
distinguishing theories)

**assumption-laden**  
(results only valid if assumptions  
hold, ignores major  
variance-explaining factors)

# Hypothesis-driven approach: distinguish two models





# Quiz

## 1 Why fMRI?

- many channels, whole-brain coverage
- noninvasive use in humans

## 2 Why pattern-information fMRI?

- The brain is multivariate.
- need population summary

## 3 Why decode fMRI?

- to test for information that might be read out from a region

## 4 Why test encoding models with fMRI?

- to test computational models of visual processing

## 5 Why should your analysis be at once data and hypothesis-driven?

- need reasonable prior assumptions to stabilise our estimates
- avoid untenable assumptions and learn from rich data