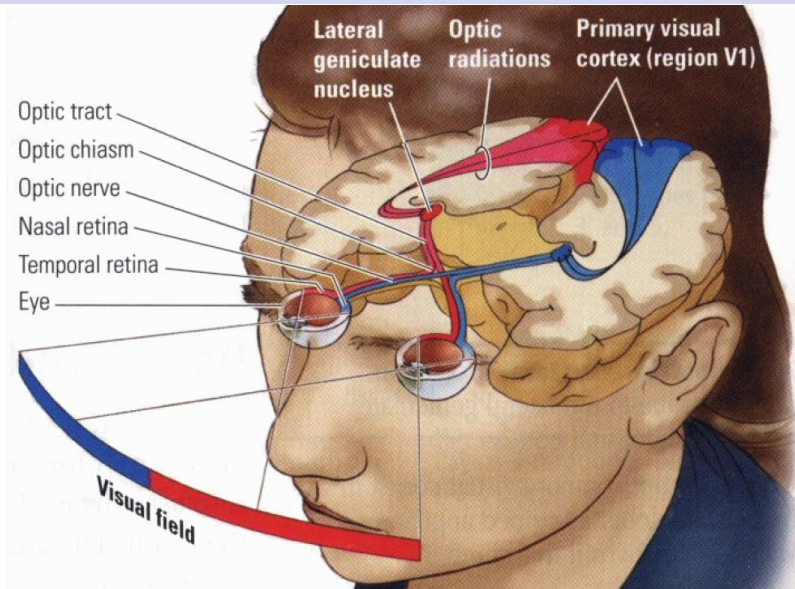
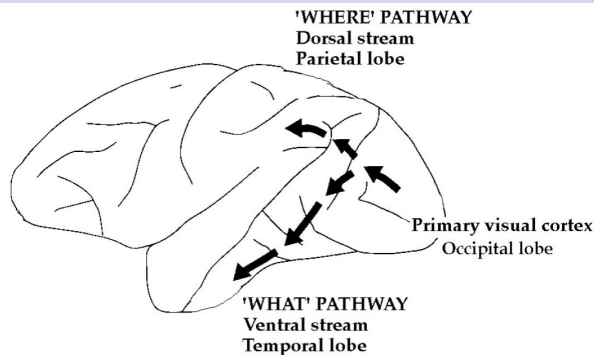


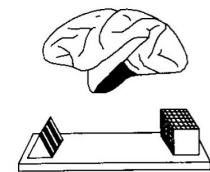
Contralateral organization of the visual system



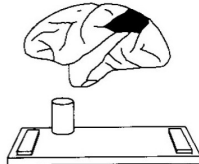
Two cortical visual systems (Ungerleider & Mishkin, 1982)



- **Dorsal (where) pathway** (upper) important of object localization
 - discriminating which food location is closer to a single object
- **Ventral (what) pathway** (lower) important for object recognition
 - discriminating between two different objects to find food
- Most deep convolutional neural networks (DCNNs) are models of the ventral pathway

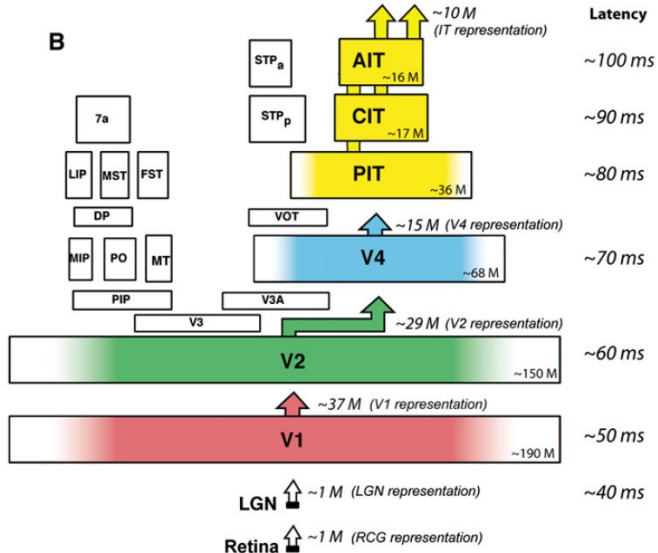
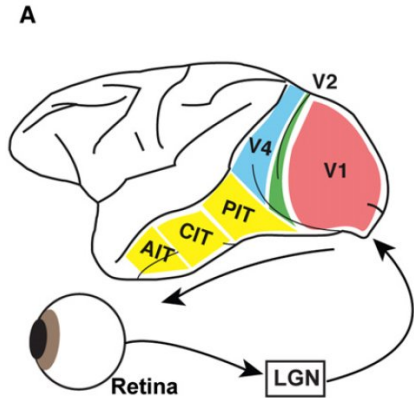


Object discrimination task
Temporal lobe lesion



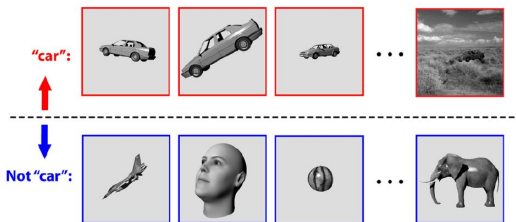
Landmark task
Parietal lobe lesion

Ventral pathway

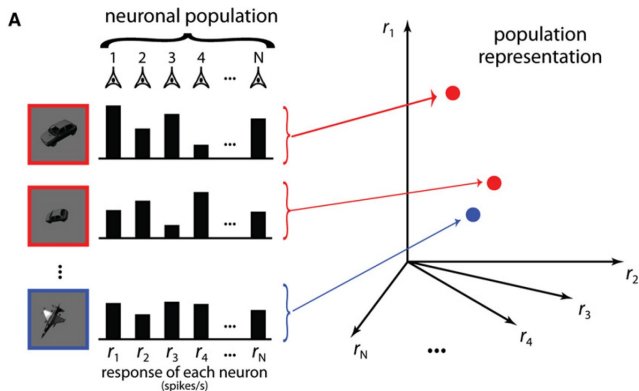


- Hierarchy of increasingly complex and spatially invariant features

Object recognition

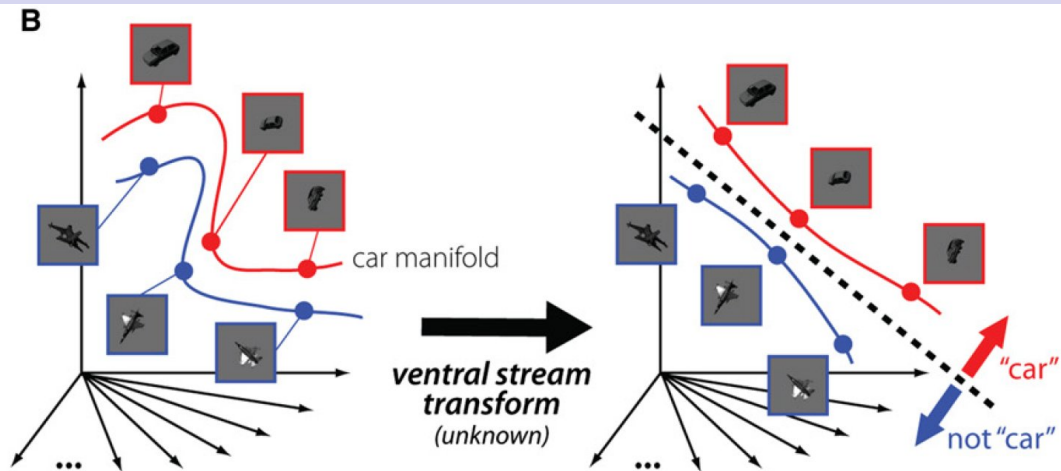


- Goal of recognition is to distinguish all images of each class/category from all other images
- Difficult because between-class image similarity can be greater than within-class similarity



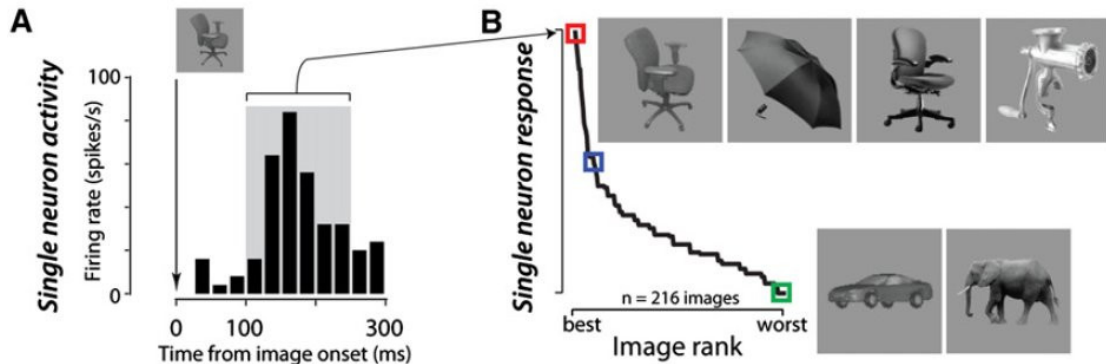
(Distributed representations in state space)

Recognition as untangling object “manifolds”



- Ventral stream must progressively alter similarity structure of representations so that different categories become linearly separable in the final layer

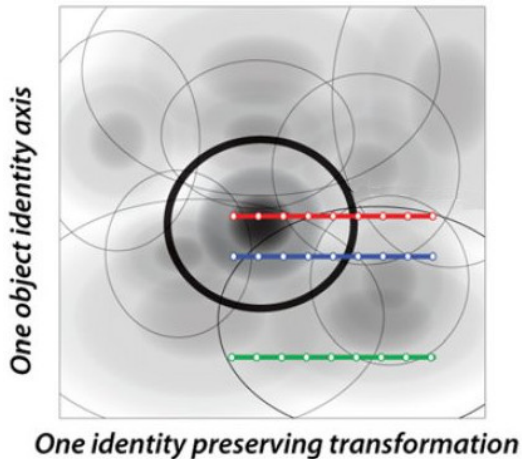
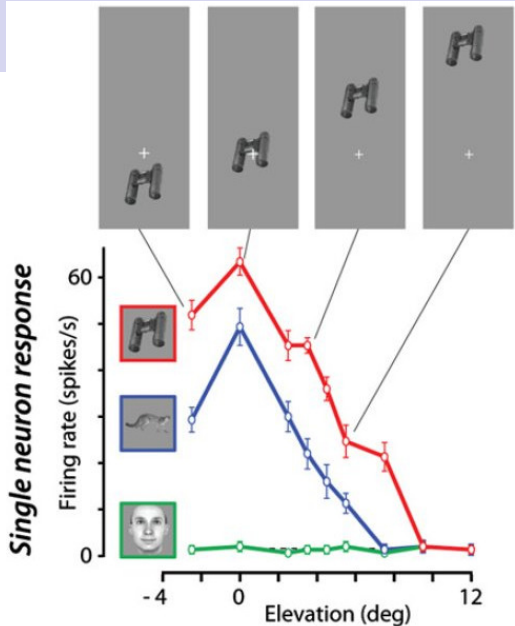
Single neuron responses in inferotemporal cortex (IT)



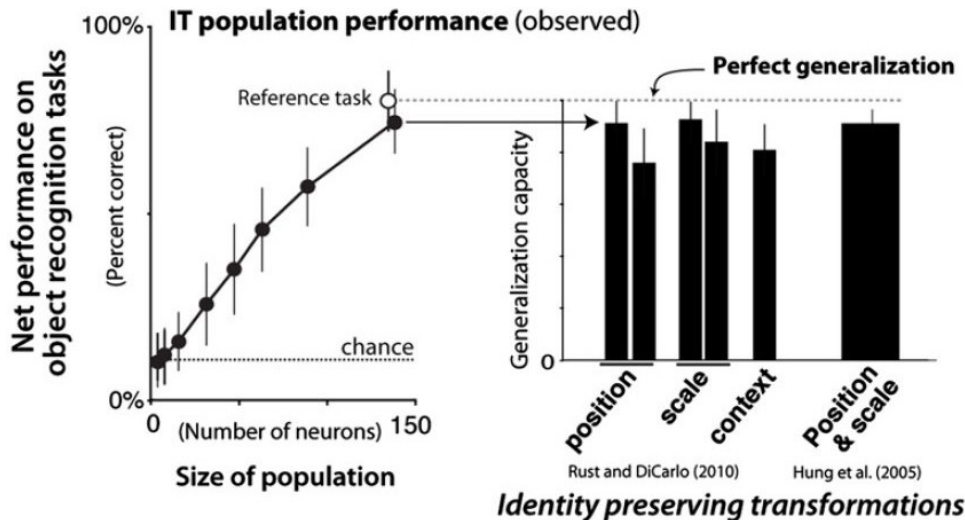
- The responses (firing rates) of individual neurons at the top layer (IT) show a degree of selectivity for different objects
 - Strong activation to few (red is best), moderate to many (blue), weak/none to most (green)

Preserving identity across viewing conditions: Single neurons

- Single neurons are **not** spatially invariant

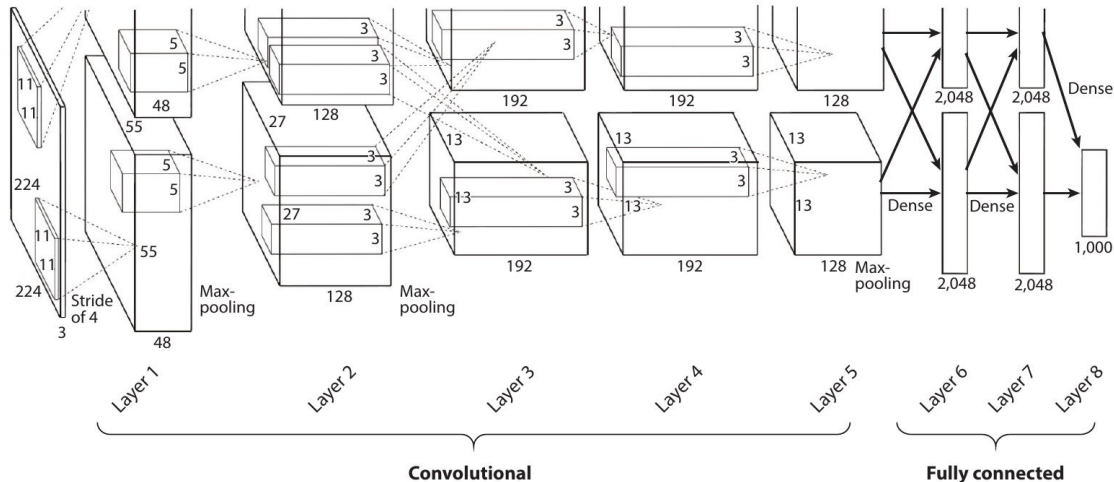


Preserving identity across viewing conditions: Population code



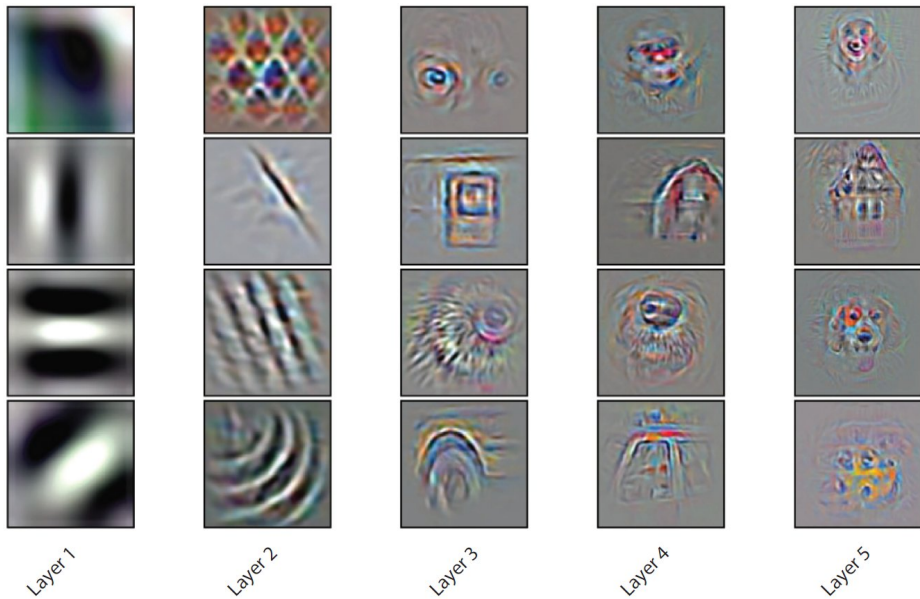
- Combining a population of IT neurons (~150) **does** support spatially invariant recognition

Deep convolutional neural networks (CNNs)

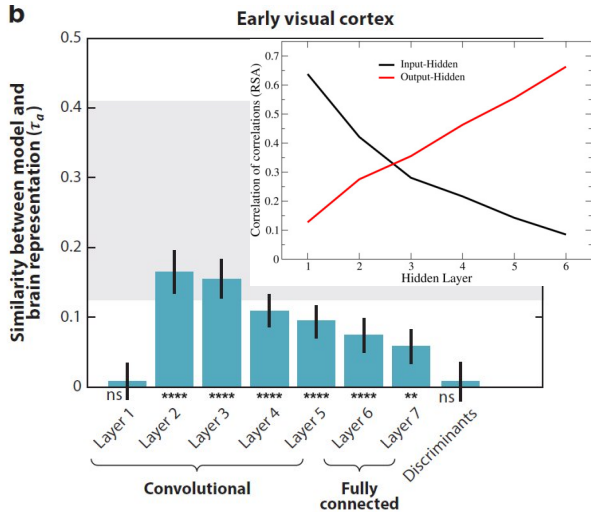
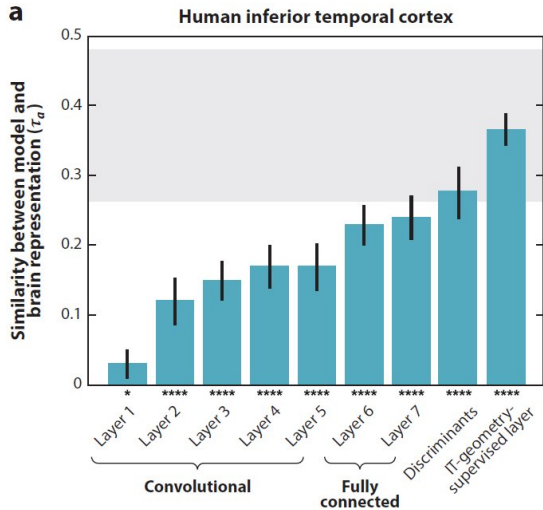


Krizhevsky, Sutskever, and Hinton (2012) “AlexNet”

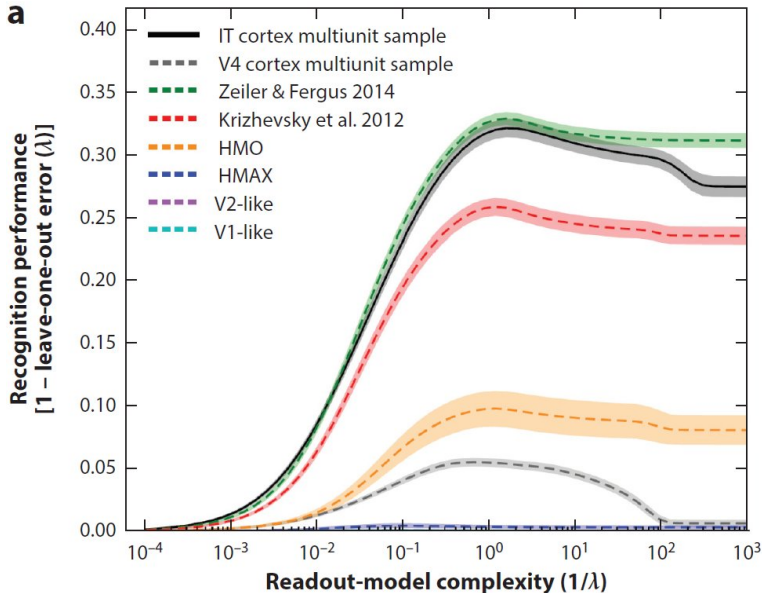
Learned features



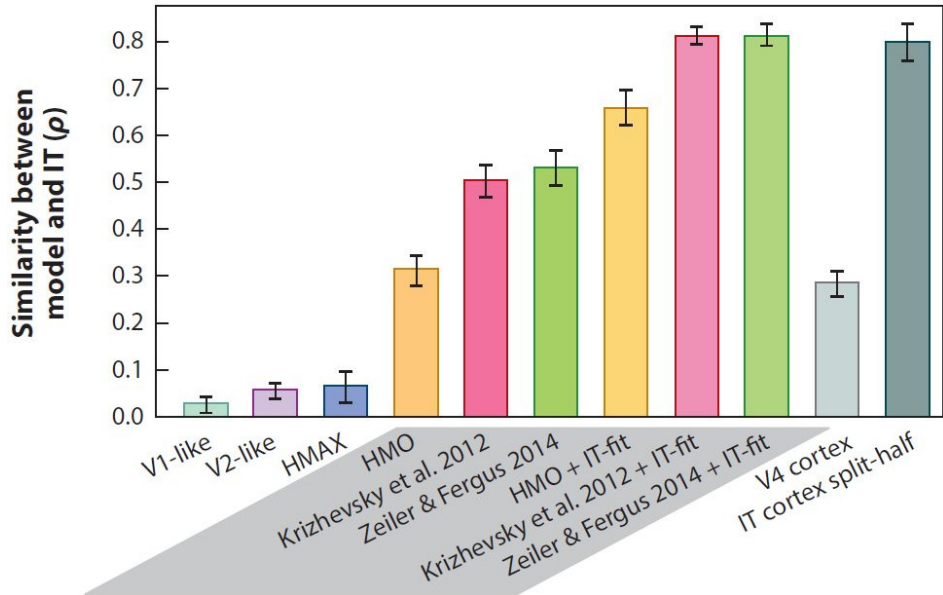
Similarity of hidden representations to neural representations



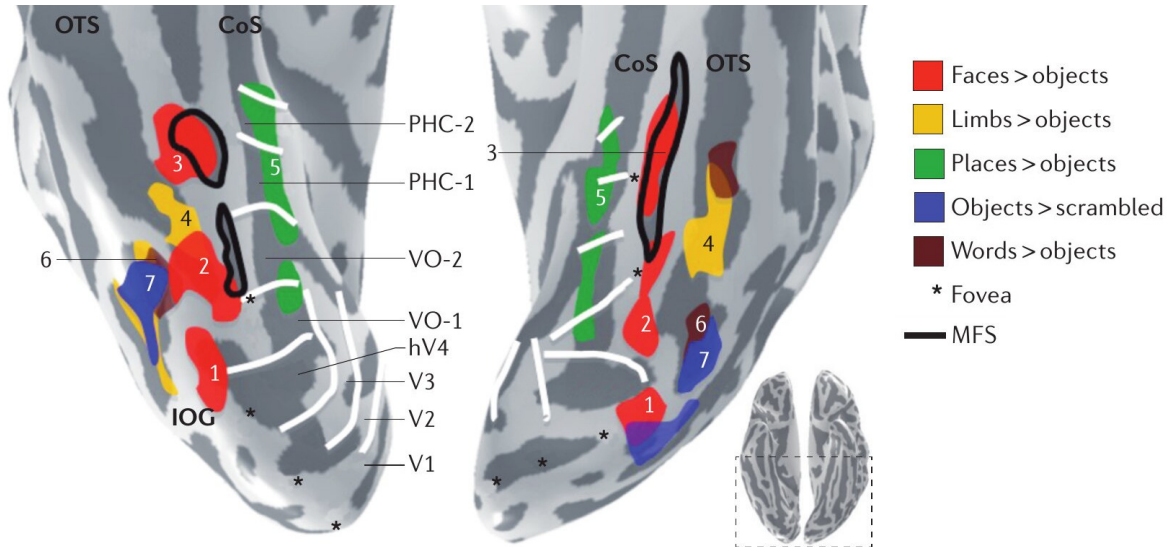
Discrimination accuracy as a function of model complexity



Similarity of representations to neural representations in IT



Domain-specific modules? (Grill-Spector & Weiner, 2014)

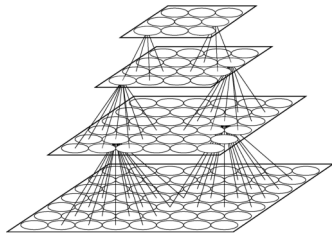


Computational principles of neural organization

Cooperation and competition among representations

Representations are hierarchically organized

The representation of information at each level, as a pattern of neural activity, **cooperates** with (i.e., mutually activates and reinforces) the representations of consistent information at lower and higher levels.



Cooperation depends on available connectivity

Connectivity is strongly constrained to minimize axon length (total volume); cooperating representations **need to be close** to each other.

Inconsistent representations compete

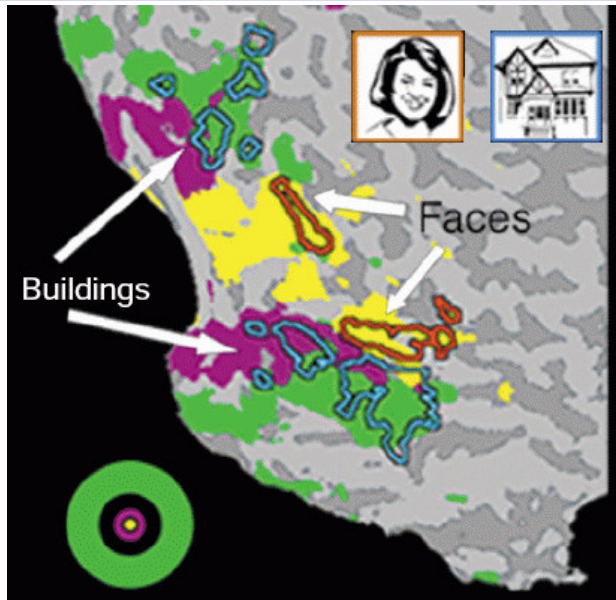
Representations of inconsistent information **compete** with each other to become active, and to become stronger through learning.

Interdependence of face and word processing (Plaut & Behrmann, 2011)

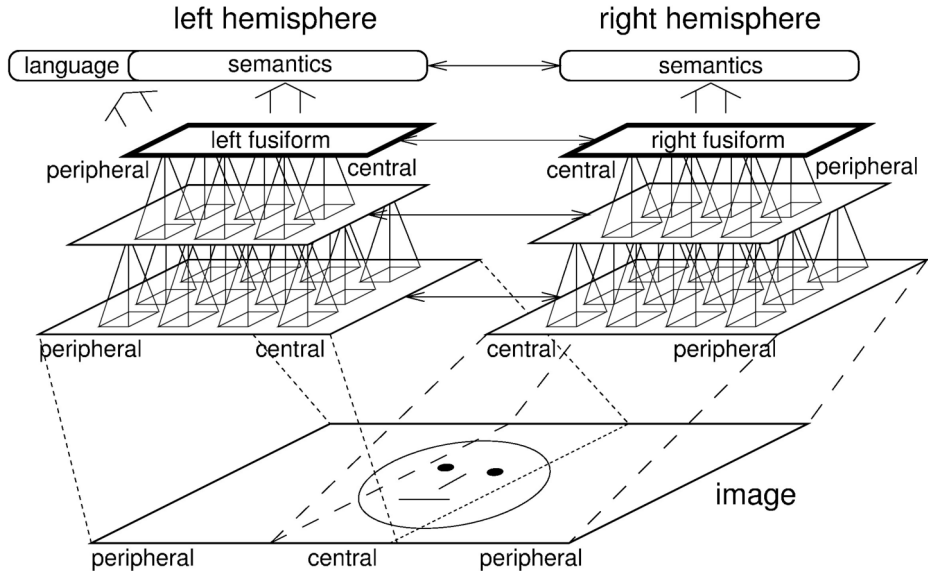
- As visual objects, faces and words are unrelated.
- However, both face and word recognition place extensive demands on **high-acuity visual information** from central vision (Malach et al.)
- Due to topographic constraints on neural organization, central visual information is **localized** in each hemisphere of the brain
- Both face and word representations need to be near central visual information to cooperate with it, but they compete with each other
- Words also need to cooperate with language-related representations (esp. phonology) which are typically left lateralized
- As a result, words become stronger in the left-hemisphere and faces become stronger in the right-hemisphere, but they are **mixed in both hemispheres** and therefore influence each other



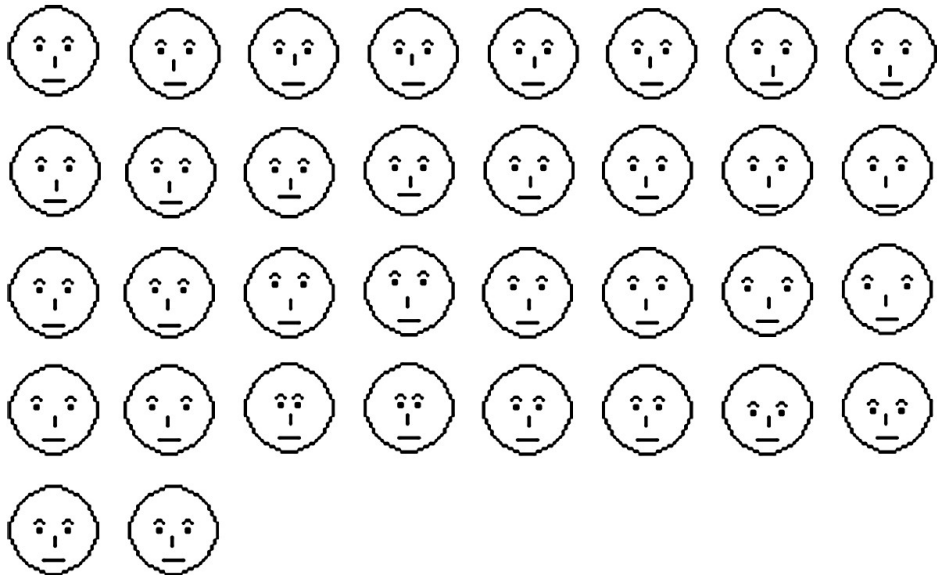
Face regions are adjacent to central visual information (Malach et al., 2002)



Simulation (Plaut & Behrmann, 2011)



Stimuli: Faces



Stimuli: Words

BAG

BED

BEG

BET

BID

BIG

BIT

BUD

BUT

DAD

DID

DIG

DIP

DOG

DOT

DUG

MUD

MET

MEN

MAT

MAP

MAN

MAD

PAD

PAN

PEN

PET

PIG

PIP

POP

POT

PUT

TAG

TAN

TAP

TEN

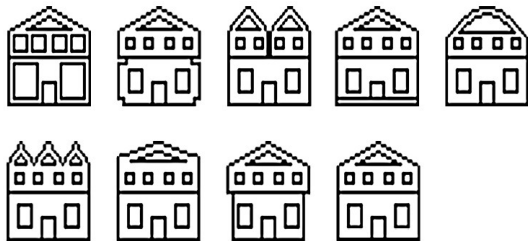
TIN

TIP

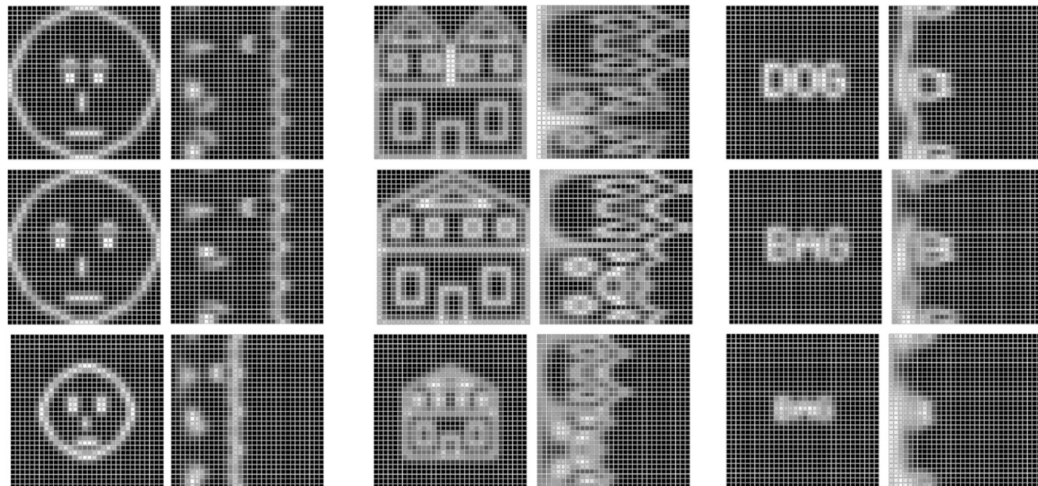
TON

TOP

Stimuli: Houses

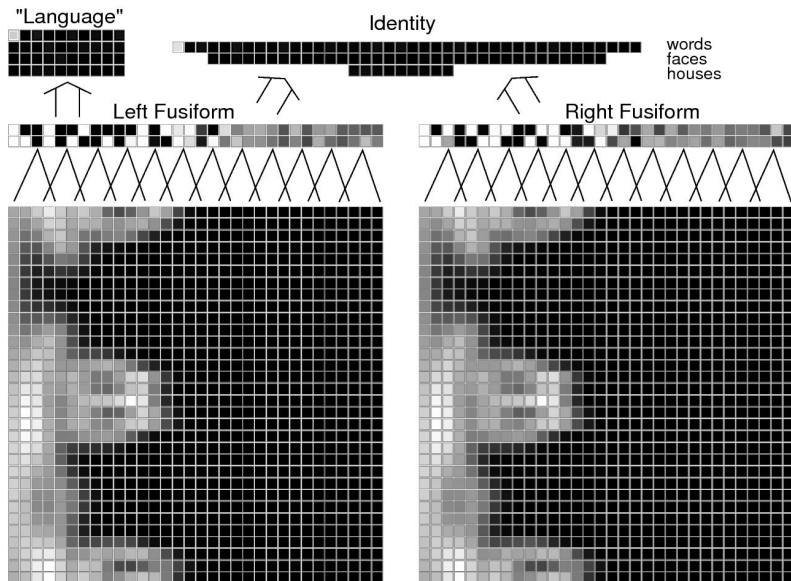


Polar coordinates; Variation in scale

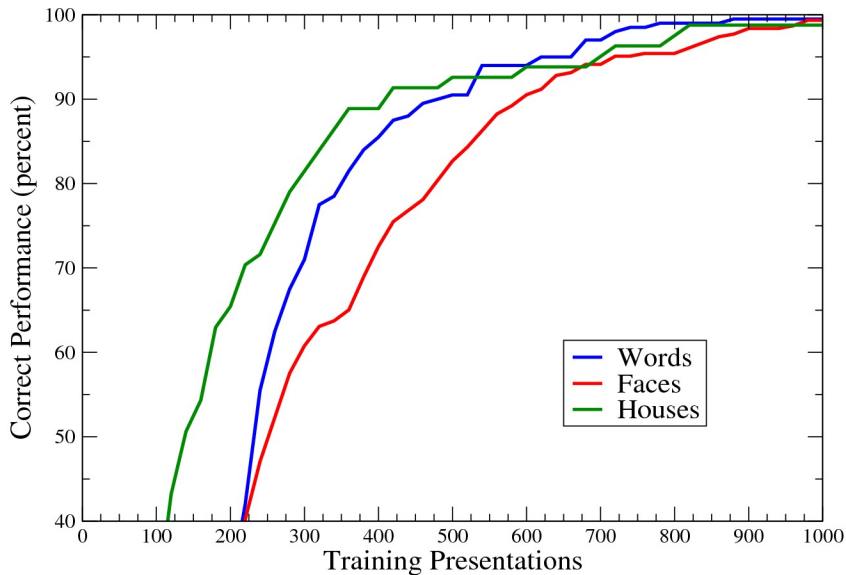


- 34 faces, 9 houses, 40 words, varying in scale
- Input in polar coordinates (eccentricity, angle)

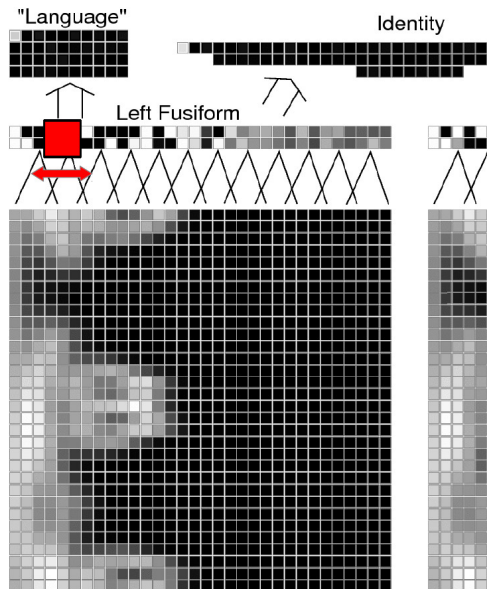
Network architecture



Acquisition



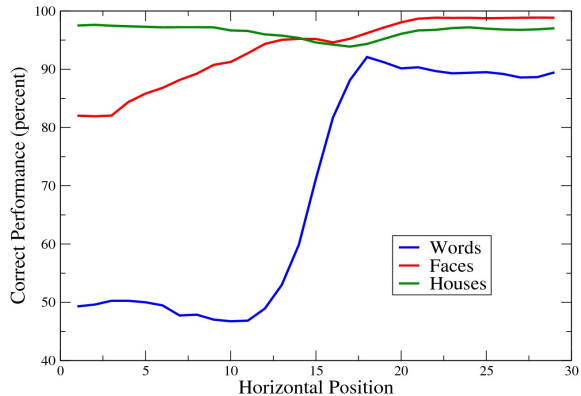
Lesioning method



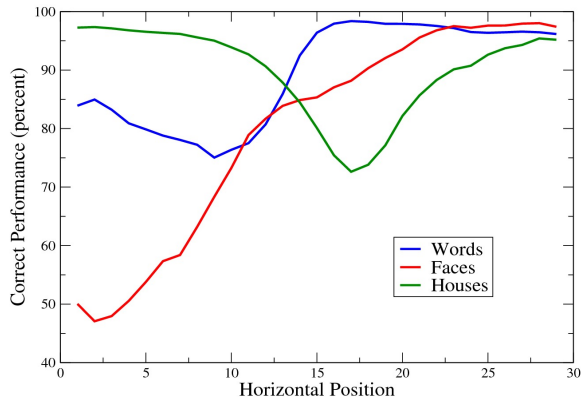
- For each horizontal position (central to peripheral) in each hemisphere, remove three adjacent columns of intermediate (fusiform) units
- Measure recognition performance on faces, words and houses (across all scales)

Lesioning results

Left Fusiform Lesions

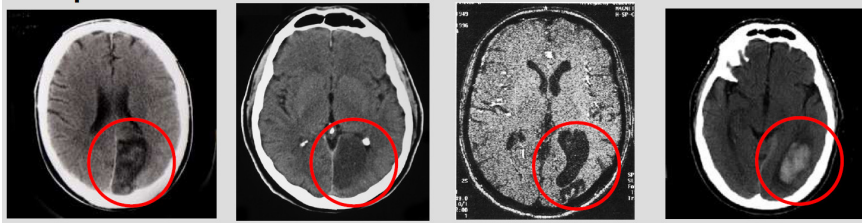


Right Fusiform Lesions



Neuropsychological data (Behrmann & Plaut, 2012)

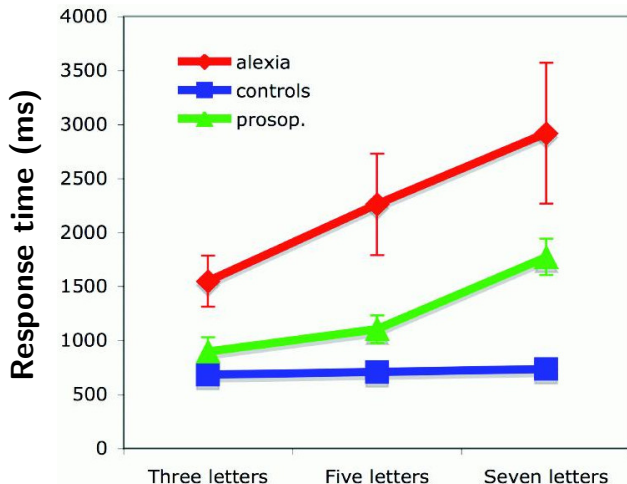
- **Pure-alexia** patients with unilateral left-hemisphere lesions (n=4)



- **Prosopagnosic** patients with unilateral right-hemisphere lesions (n=3)

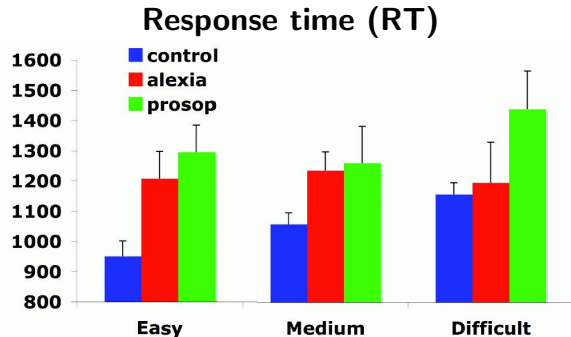
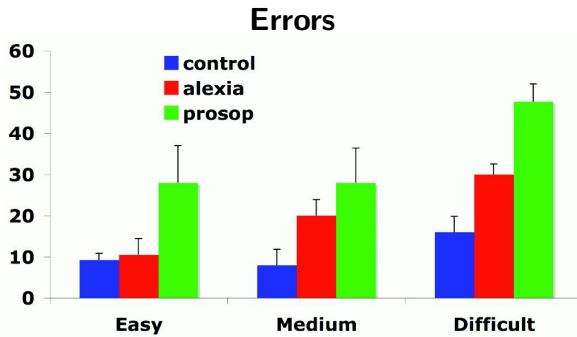


Word processing: Lexical decision



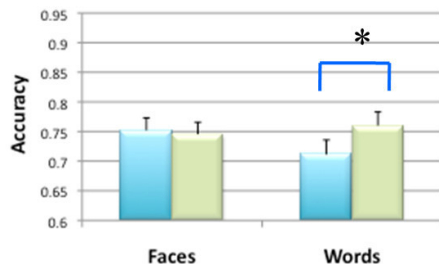
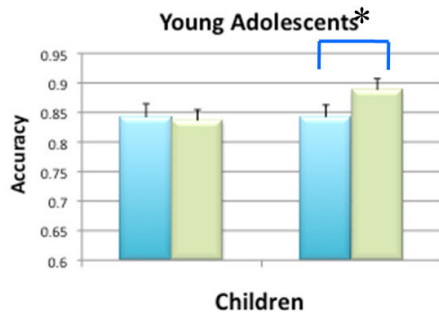
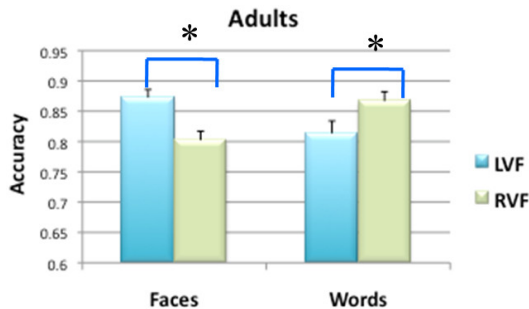
- Unlike normal controls (blue), pure alexic patients (red) read longer words much more slowly than shorter words (“letter-by-letter reading”)
- Prosopagnosic patients (green) are mildly pure alexic—show weaker but still abnormal length effect

Face processing: Same-different judgements

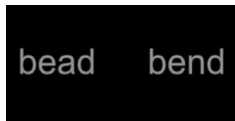
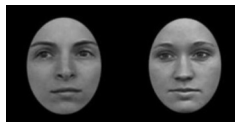


- Prosopagnosic patients (green) are much worse than controls (blue) at same-difference judgements with faces
- Pure alexic patients (red) are mildly prosopagnosic (worse than controls, better than prosopagnosics)

Word lateralization drives face lateralization (Dundas et al., 2012)

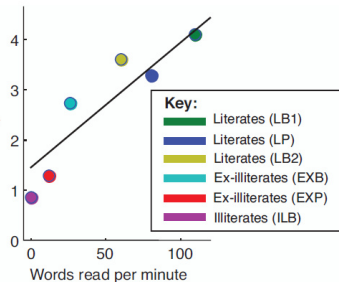
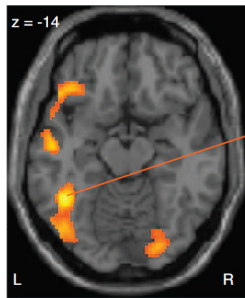


Sequential same-different judgment where second stimulus is presented to the left or right of fixation (LVF or RVF)



Effects of literacy on VWFA activation (Dehaene et al., 2010)

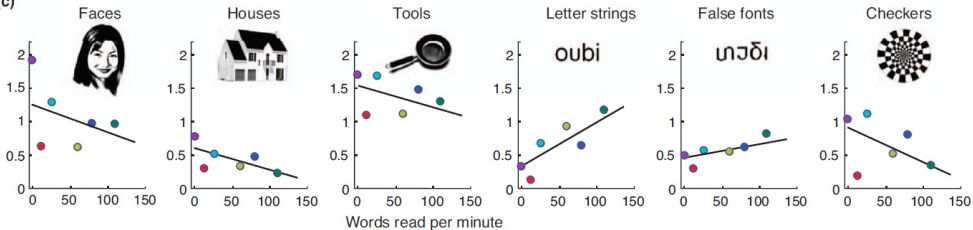
(a) Written sentences



(b) Letter strings



(c)

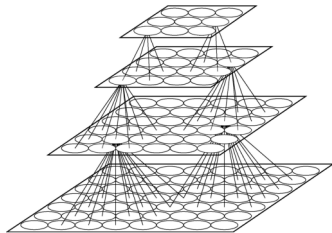


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