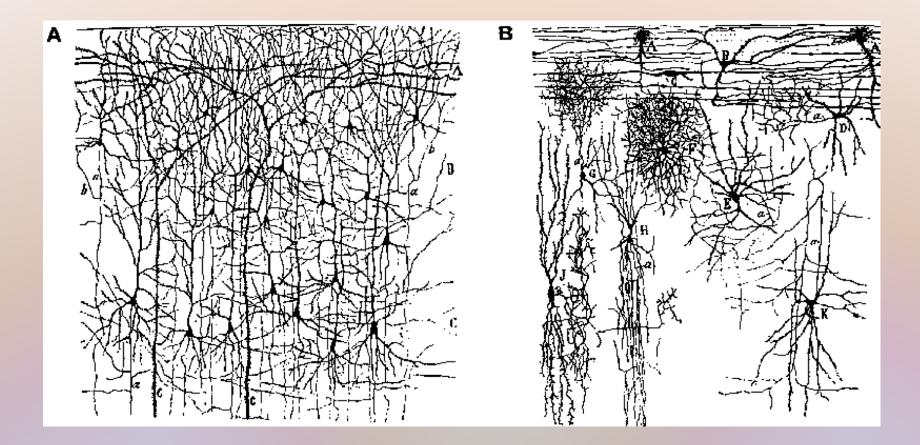
3. Networks

Networks

- 1. Biology of Neocortex ("cortex")
- 2. Categorization and Distributed Reps
- 3. Bidirectional Excitation and Attractors
- 4. Inhibitory Competition and Activity Regulation

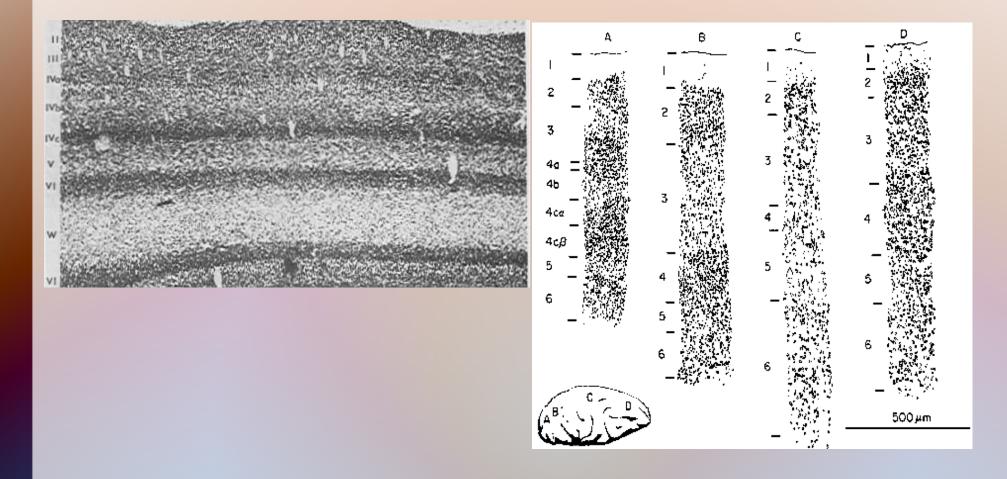
Neurons: Excitatory and Inhibitory



Excitatory = main info processing, long-range connections Inhibitory = local, activity regulation and competition

(slide < O'Reilly)

The 6 Cortical Layers



(slide < O'Reilly)

COSC 421/521

4

Functions of Layers

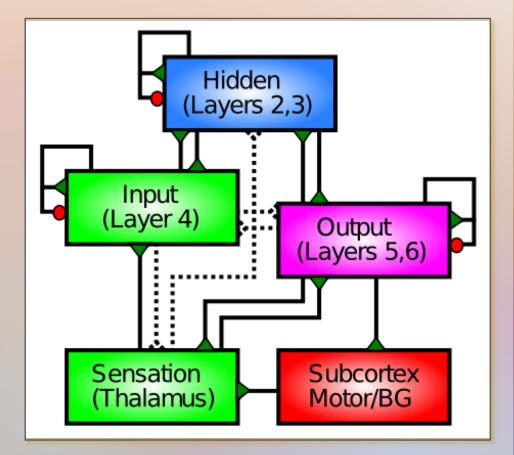
Input

layer 4 from sensation or other areas Hidden

layers 2 & 3

Output

layers 5 & 6 to motor systems or other areas



(fig. < O'Reilly, Comp. Cog. Neurosci.)

Connection Directions

Feedforward

Feedforward

• from Hidden in lower area to Input in higher area

Feedback

• from Hidden & Output in higher area to Hidden & Output in lower area

Lateral

• from Hidden and Output to all three layers in same area

Bidirectionality

Terminals Activation Flow Activation Flow Source Neurons Lateral Hidden (2,3) Input (4) Output (5,6) Activation Flow (fig. < O'Reilly, *Comp*. COSC 421/521 Cog. Neurosci.)

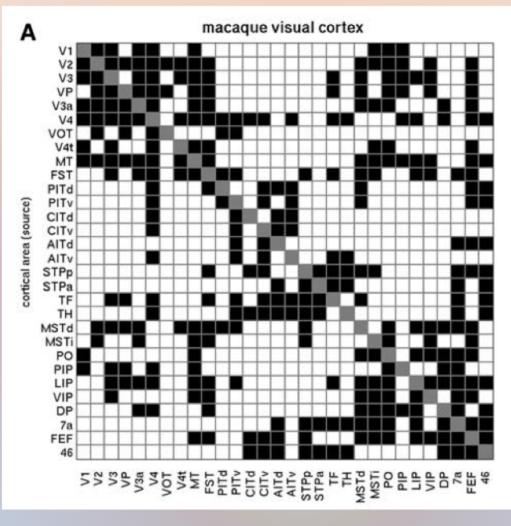
Axon

Feedback

6

• pervasive

Bidirectional Symmetry



(slide < O'Reilly)

Biology \Rightarrow Function

- Feedforward excitation = categorization of inputs

 larger patterns, more invariant w.r.t instances & space

 Feedback excitation = attractor dynamics

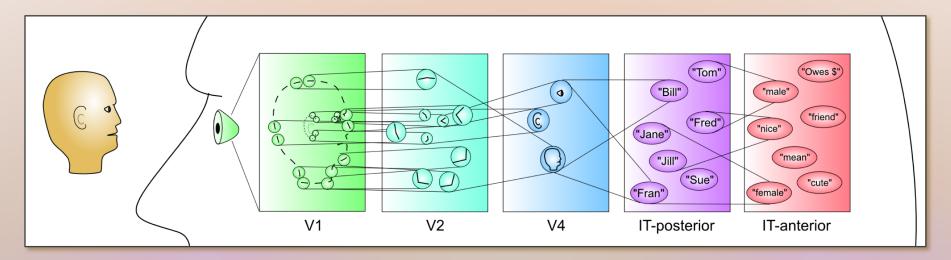
 ambiguity resolution & constraint satisfaction
- Lateral inhibition = competition, activity regulation
 sharpens response

Ambiguity Resolution



COSC 421/521 (fig. < O'Reilly, *Comp. Cog. Neurosci.*)

Hierarchical Categorical Representations



- Successive layers of neural detectors
- Progressively more abstract

COSC 421/521 (fig. < O'Reilly, *Comp. Cog. Neurosci.*)

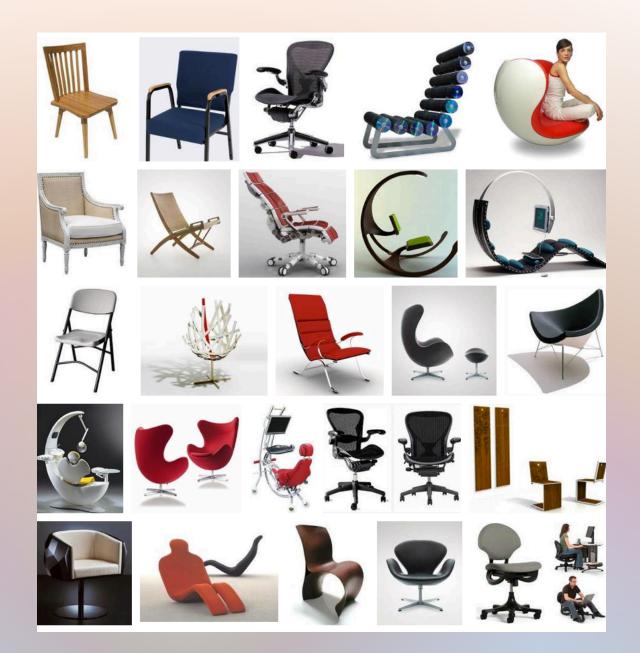
Getting the right representation is key

- Two men are dead in a cabin in the woods.
- The cabin itself is not burned, but the forest all around is burned to cinders.
- How did the men die?

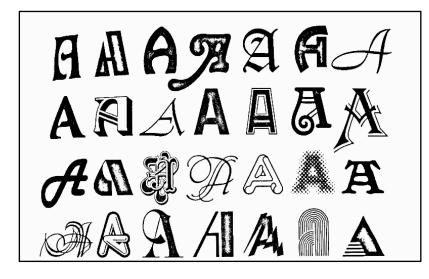
The Chair Category

How would you define "chair"?

Socrates asks, "what is that common quality, which is the same in all these cases, and which is called courage?" (*Laches* 191e)



The central problem of AI is: What are *a* and *i*? — Douglas Hofstadter





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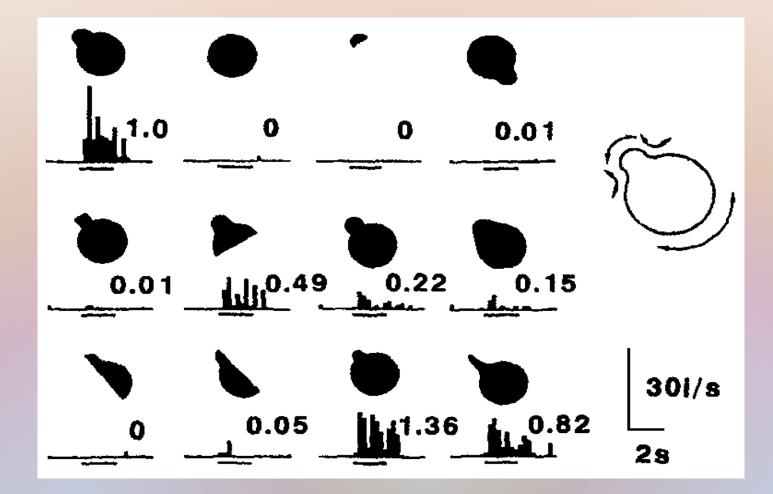
Categories: A Philosophical Problem

- A long-standing problem:
 - Socrates (d. 399 BCE) says, "that which we know we must surely be able to tell." (*Laches* 190c)
 - Must knowledge be encoded in language-like structures?
- What makes a mental categorization accurate? Is there something "real" about a "chair"?
- Stereotypes are mental categories...
- Can you encode multiple categories at the same time?

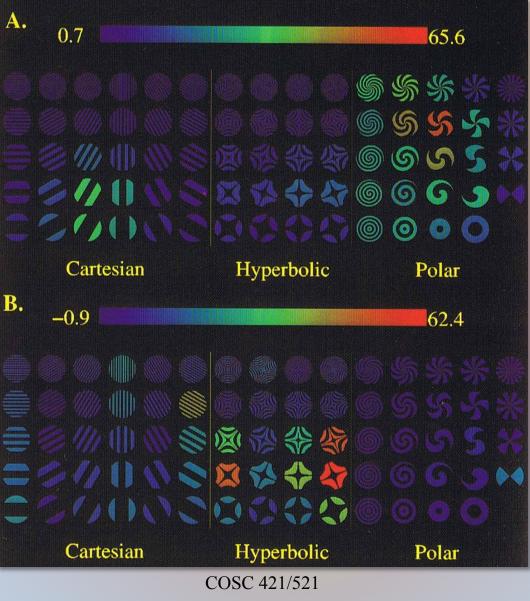
Distributed Representations

- Let a 1,000 categories bloom... You've got the room in your head (billions of neurons)
- Each neuron can respond to multiple things (graded similarity)
- And each thing activates many neurons (who knows what is going to be relevant this time?)

Graded Responses

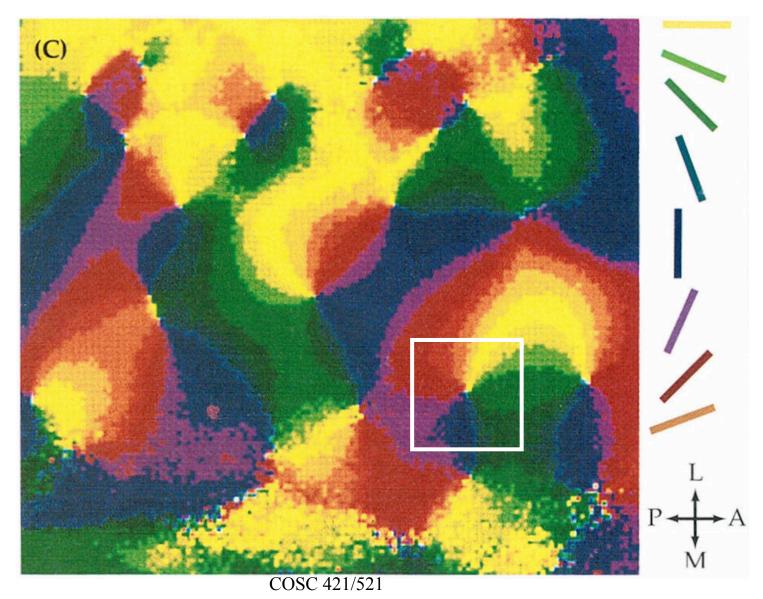


Cell Responses in V4



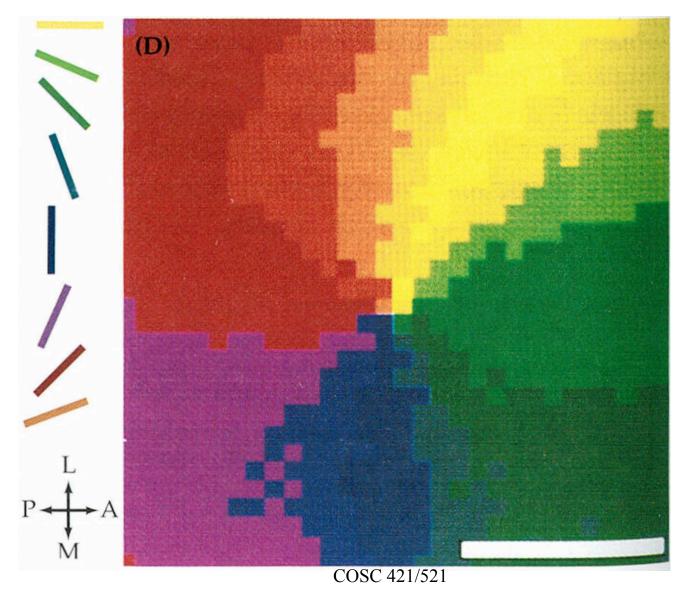
(fig. < Clark, *Being There*, 1997)

Orientation Columns



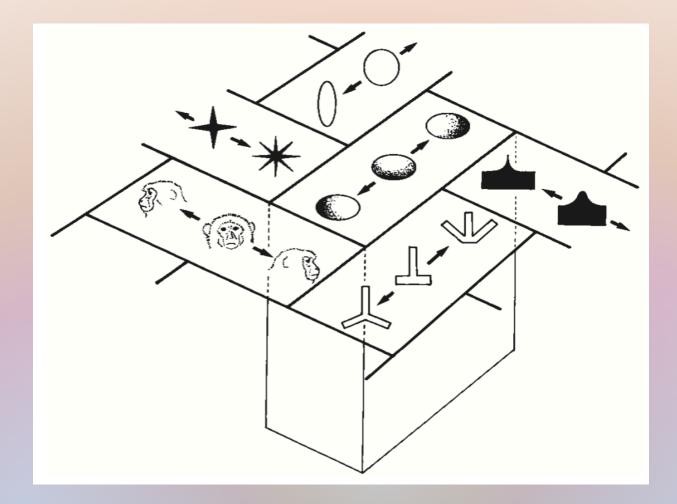
(fig. < Nicholls & al., Neur. to Brain)

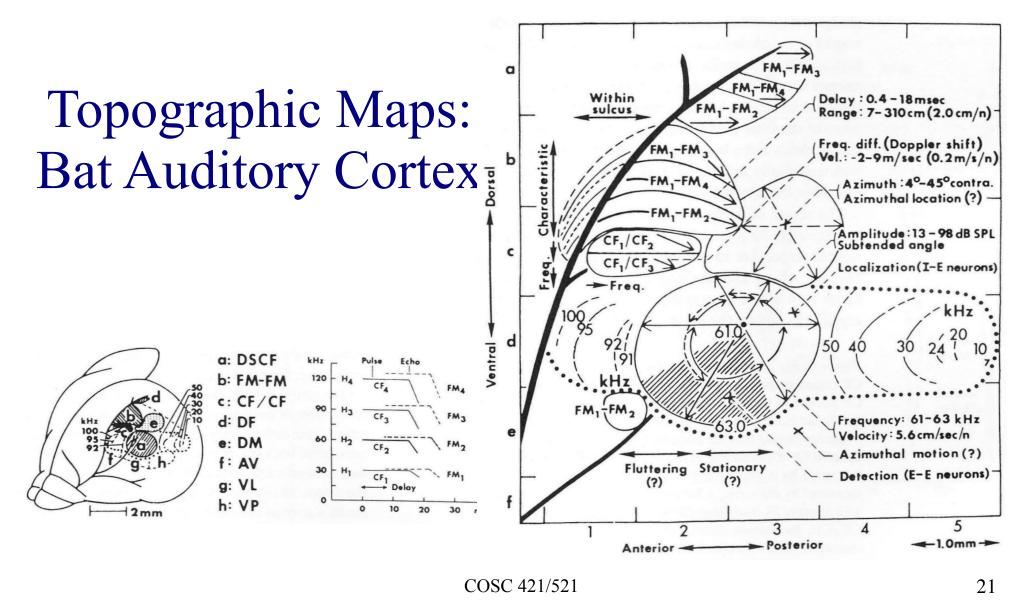
Orientation Columns



(fig. < Nicholls & al., Neur. to Brain)

Topographic Organization

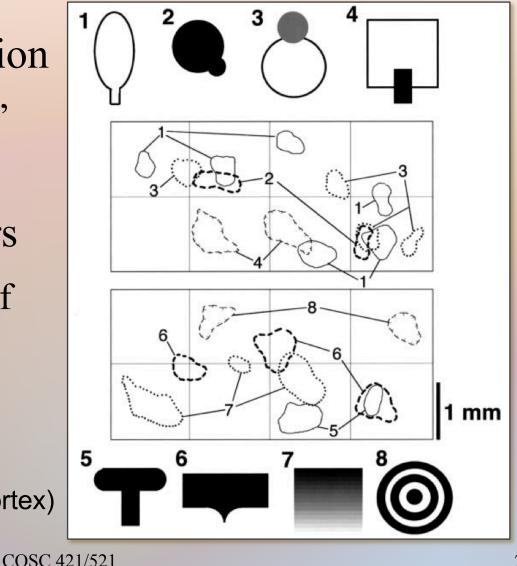




(figs. from Suga, 1985)

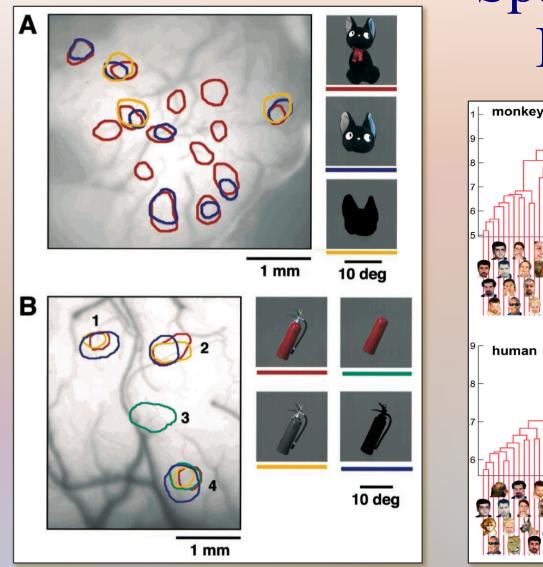
Sparse Distributed Representation

- Localist representation
 - "grandmother cells"— unlikely in brain
- K-out-of-N detectors
 - typically 15–25% of neurons active
- Approximate orthogonality

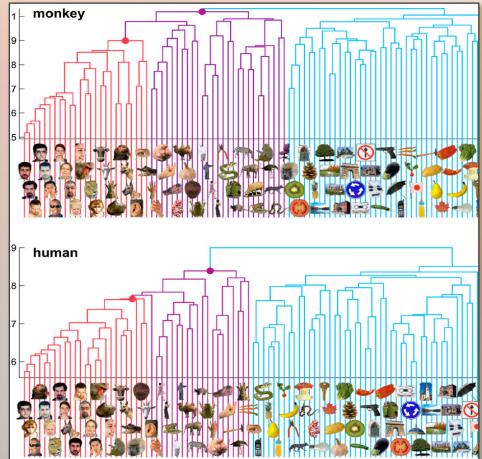


(monkey IT cortex)

(fig. < O'Reilly, *Comp. Cog. Neurosci.*, from Tanaka, 2003)



Sparse Distributed Representations

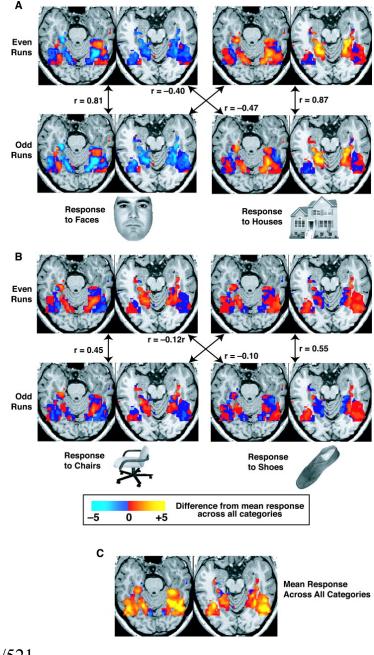


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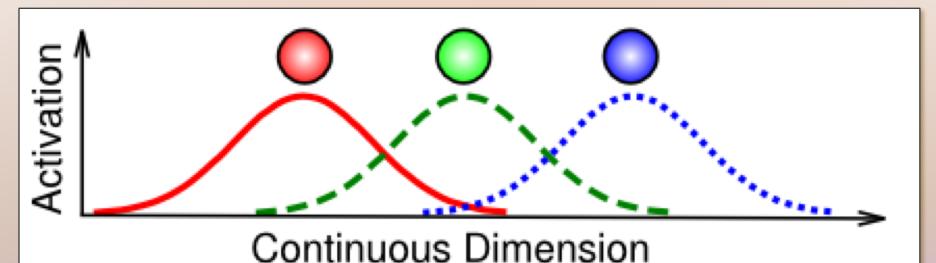
(figs. < O'Reilly, Comp. Cog. Neurosci., from Tanaka 2003 and Kriegeskorte et al. 2008)

Not Just Monkeys

- Maps of neural activity in the human brain in response to different visual input stimuli
- (as shown faces, houses, chairs, shoes)
- Recorded using functional magnetic resonance imaging (fMRI)
- There is a high level of overlap in neural activity across these different stimuli, in addition to some level of specialization
- This is the hallmark of a distributed representation



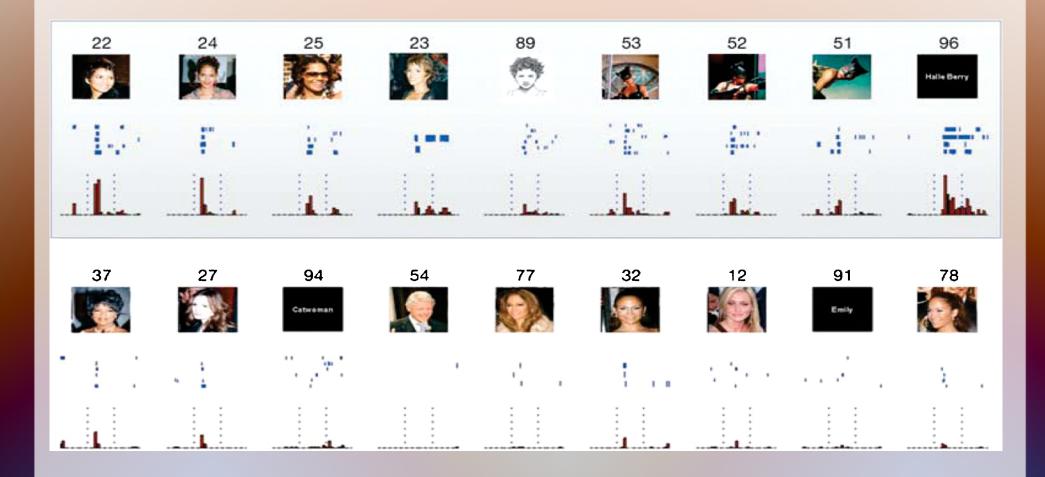




- Broadly-tuned receptive fields
- Population-coding of precise values
- Common throughout sensory and motor areas

COSC 421/521 (fig. < O'Reilly, *Comp. Cog. Neurosci.*)

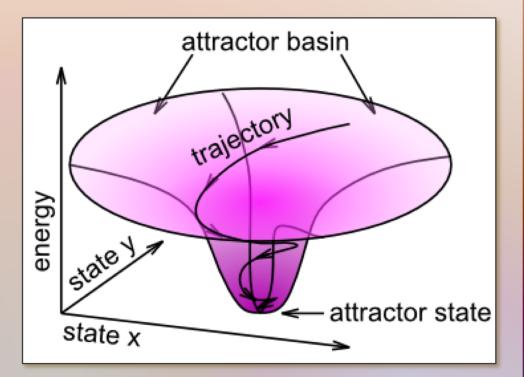
Localist Representations?



<u>emergent</u> demonstration: Face Categorization I

Bidirectional Excitation

- Functions
 - recognition
 - top-down imagery
 - ambiguity resolution
 - pattern completion
- Attractor dynamics
 - convergence on good representation
 - energy vs. harmony



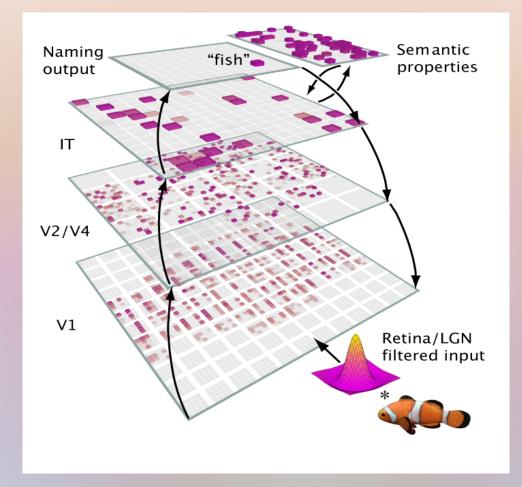
(fig. < O'Reilly, Comp. Cog. Neurosci.)

What Are These?

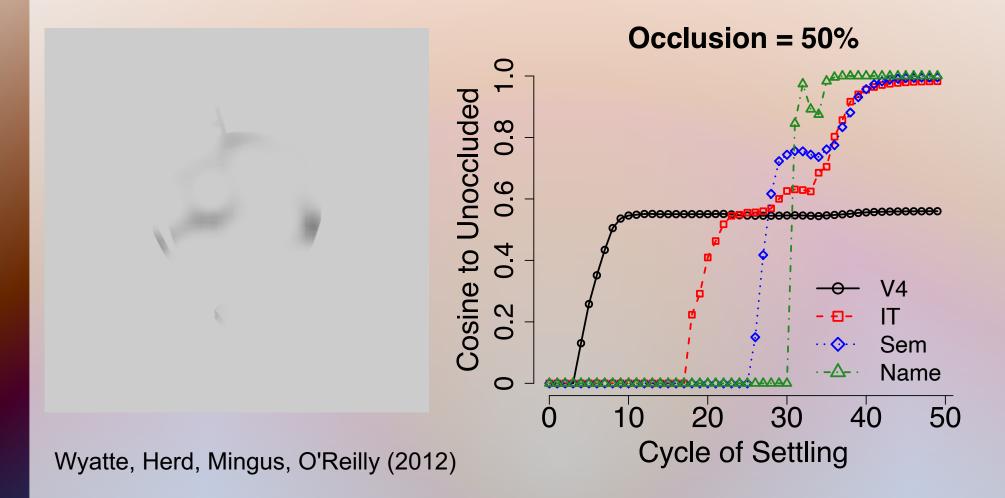


(slide < O'Reilly)

A Big Network Model...



Bidirectional Dynamics



<u>emergent</u> demonstration: Face Categorization II

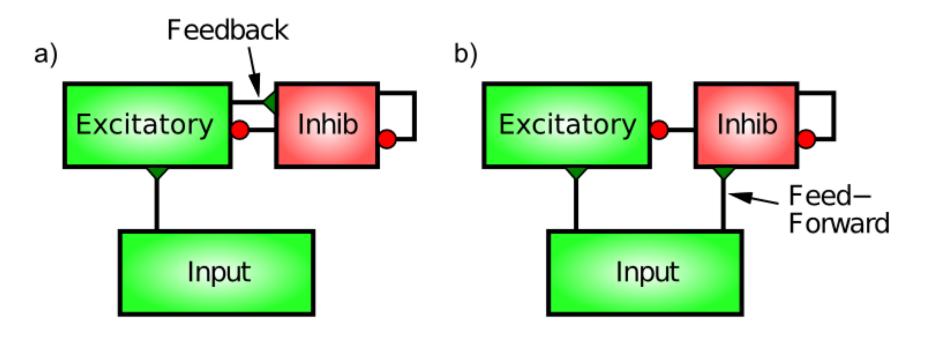
emergent demonstration: Cats and Dogs

emergent demonstration: Necker Cube

Inhibitory Competition and Activity Regulation

- Activity regulation
- Selective attention
- Competition
- Sparse distributed representation

Activity Regulation

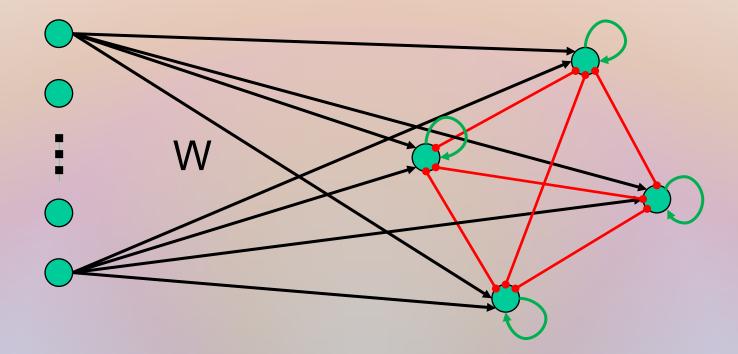


- Feedback: reactive, reflects actual level of activity, robust, responsive, may be unstable
- Feedforward: anticipatory, limits feedback oscillation, slow, brittle
- Work well together

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(fig. < O'Reilly, Comp. Cog. Neurosci.)

Competitive Network



Competitive Classification

- two layers
- first layer of detectors
- second is self-reinforcing, mutually inhibitory
- "winner takes all" dynamics
- implements nearest neighbor classification

FFFB Inhibition Approximation

- Approximates total effect of all inhibition in a layer
- Inhibition determined by feedforward and feedback terms: $g_i(t) = gi[FF(t) + FB(t)]$
- FF term is excess average input over set point: $FF(t) = ff[\langle \eta \rangle - ff0]^+$ where $\langle \eta \rangle = n^{-1} \sum_{i=1}^{n} \eta_i$ is average input
- FB term varies with average activity: $\dot{FB}(t) = dt[fb\langle y \rangle - FB(t)]$ where $\langle y \rangle = n^{-1} \sum_{i=1}^{n} y_i$ is average activity
- Will stabilize with $FB(t) = fb\langle y \rangle$

<u>emergent</u> demonstration: Inhibition