

Neural Dynamics For Embodied Cognition

Lecturers:

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Survey

■ Session 1: Foundations

- Neural dynamics/neural fields [Daniel Sabinasz]
- Introduction to Cedar/Instabilities in DFT [Raul Grieben]

■ Session 2: Dimensions/Binding [Raul Grieben]

- Cedar architecture: visual search

Survey

■ Session 3: Grounded Cognition [Daniel Sabinasz]

- Cedar architecture: relational grounding

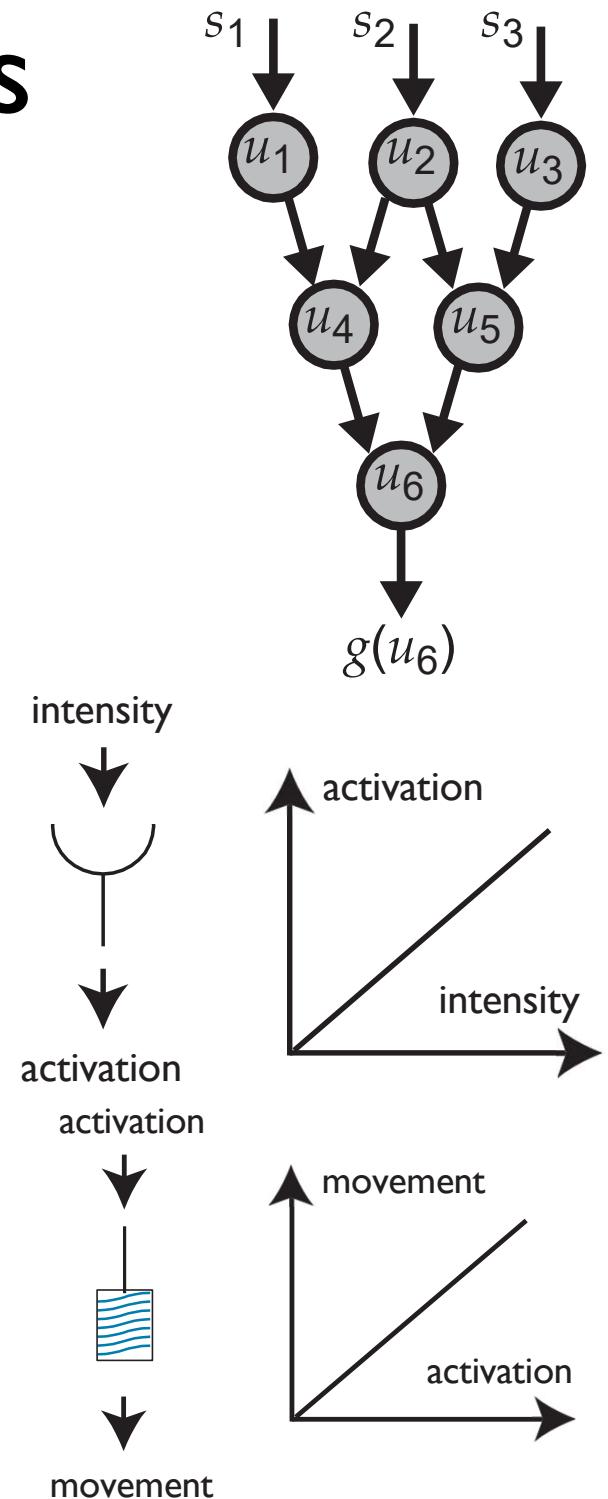
■ Session 4: Sequence generation

- Sequence generation/Embedding DFT [Raul Grieben]
- Cedar architecture sequence generation [Daniel Sabinasz]

- Neural fields: dimensions
- Binding
- Visual search

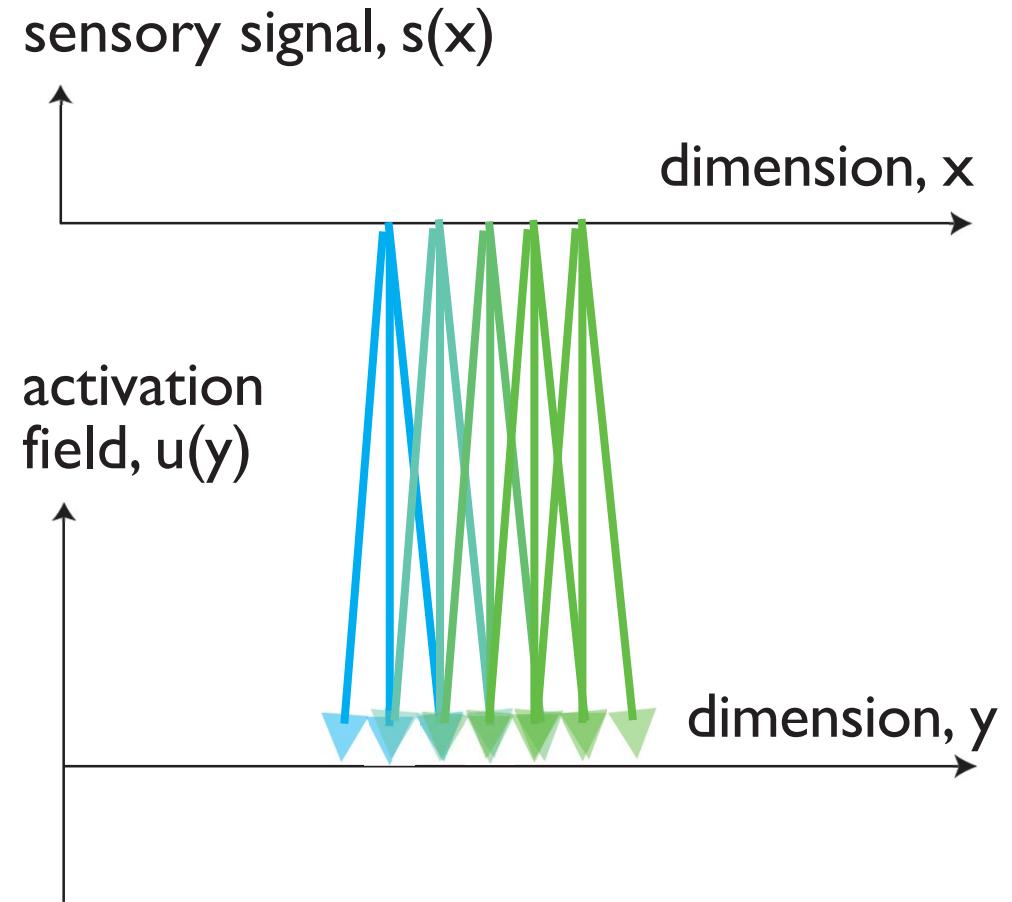
Neural networks

- forward connectivity
determines “what a neuron stands for” = **space code** (or labelled line code)
- while the activation level may “stand for” intensities = **rate code**
- generic neural networks combine both codes



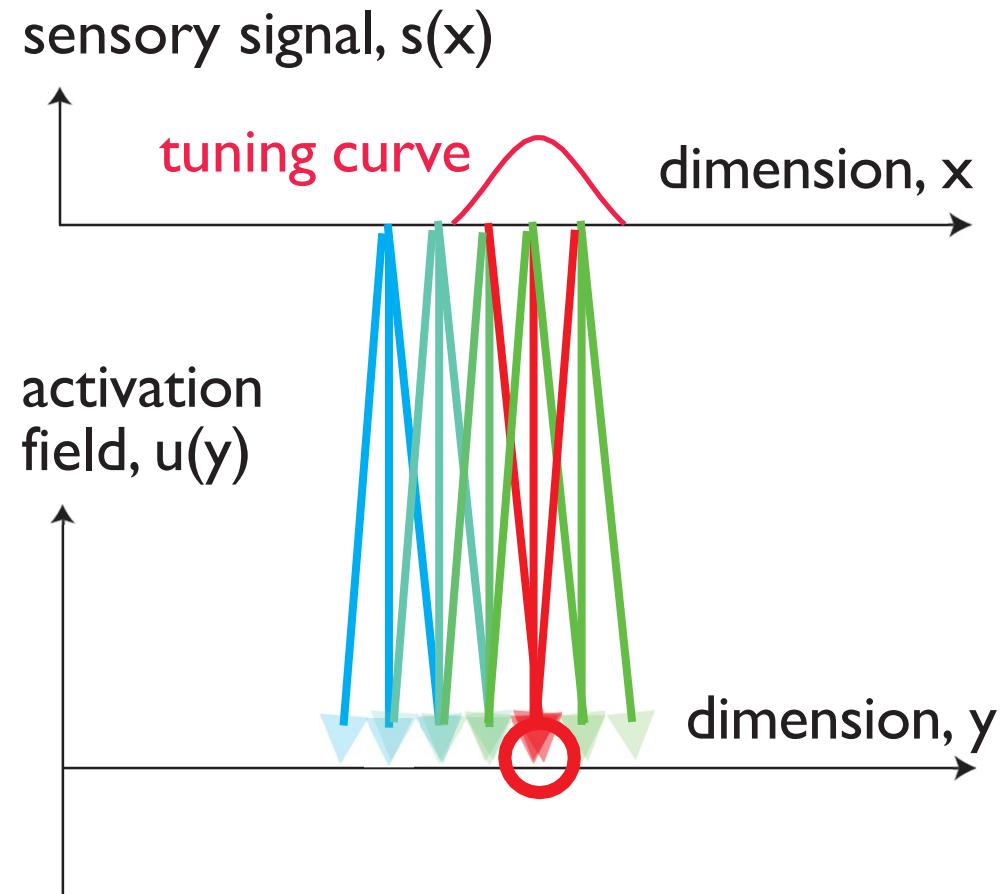
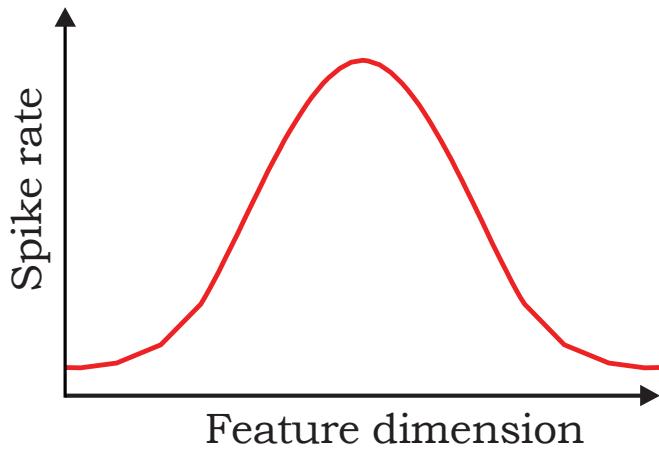
Neural fields

■ forward connectivity
from the sensory surface
extracts perceptual
feature dimensions



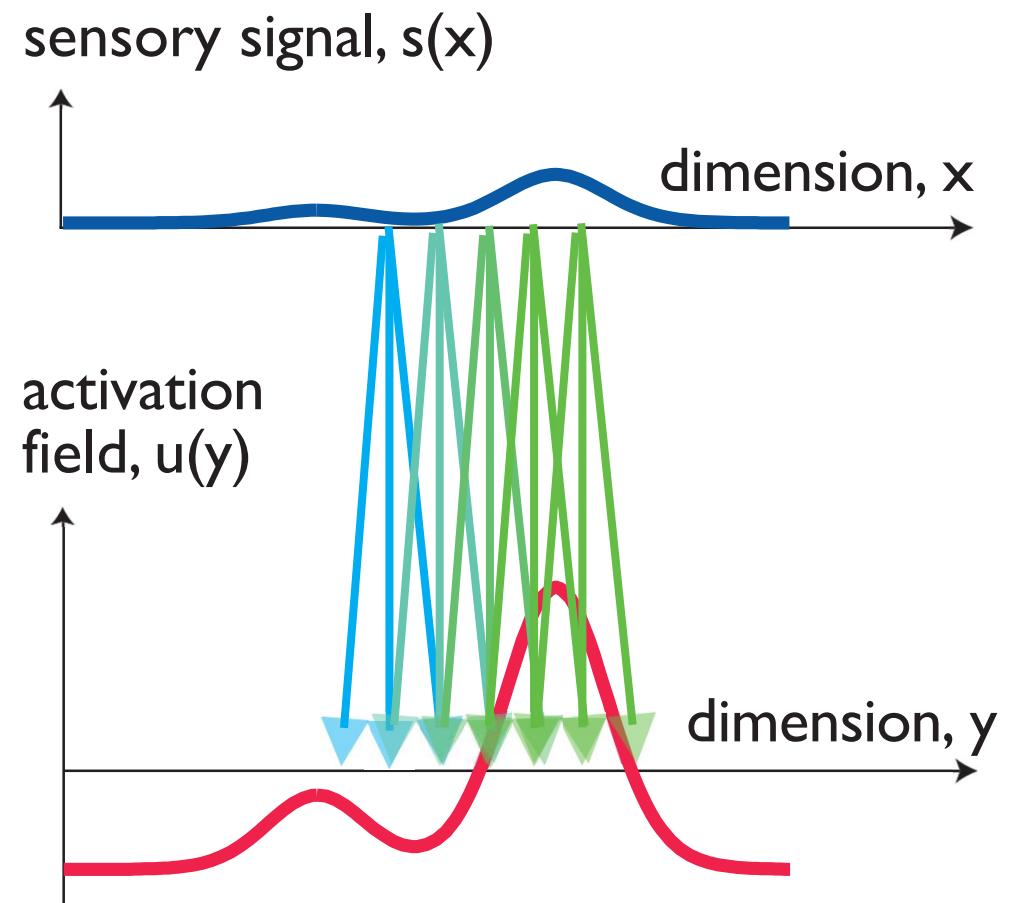
Neural fields

- as described by tuning curves or receptive fields



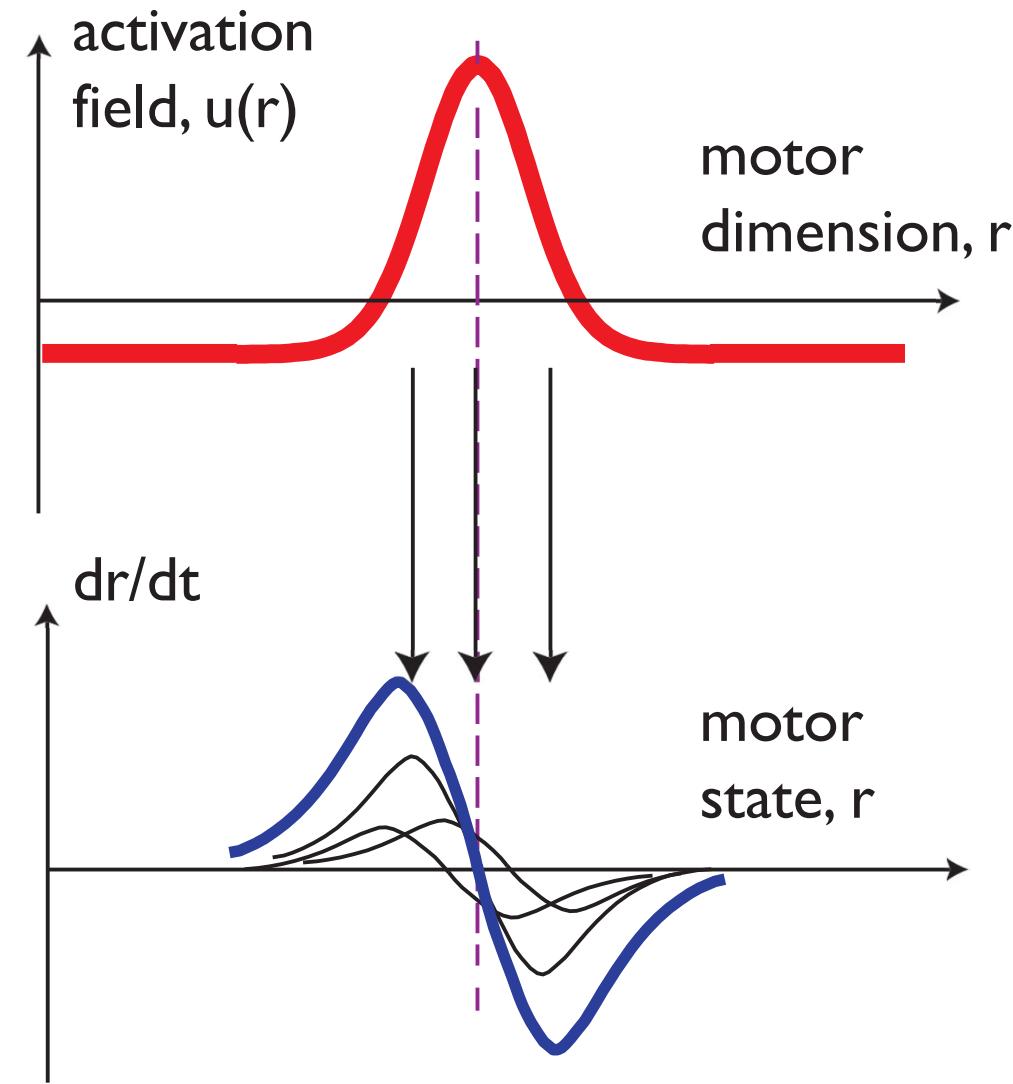
Neural fields

- => **neural map** from sensory surface to feature dimension
- neglect the sampling by individual neurons => **activation field**



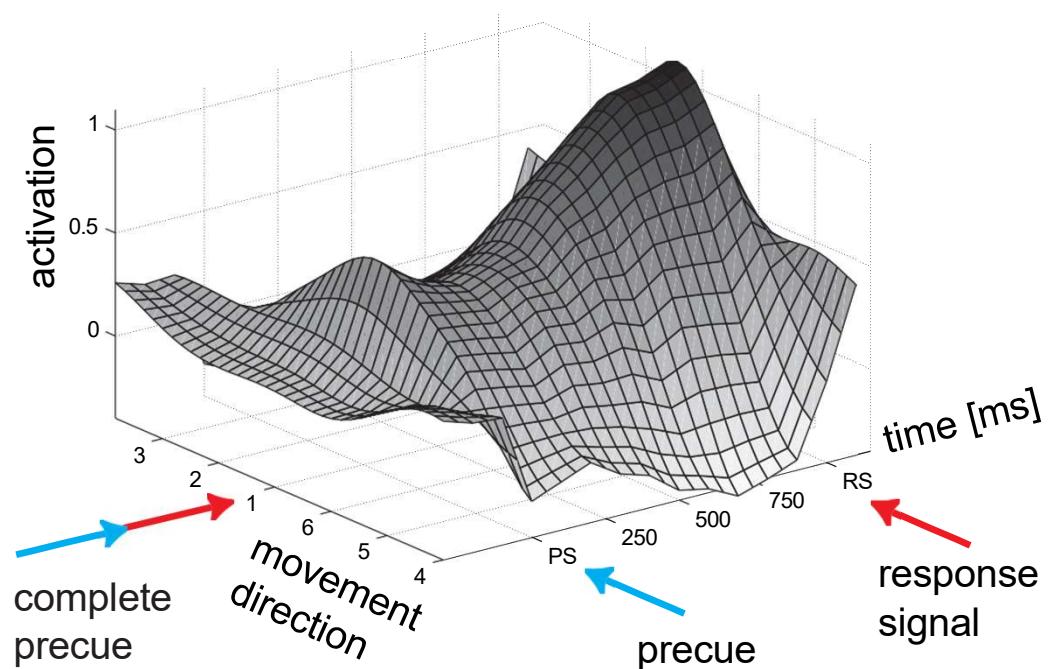
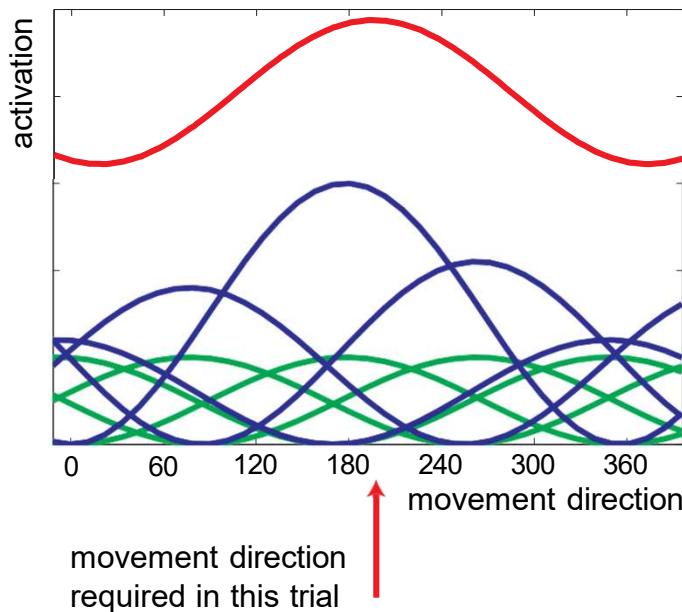
Neural fields

- analogous for projection onto to motor surfaces...
- which actually involves behavioral dynamics (e.g., through neural oscillators and peripheral reflex loops)



Distribution of Population Activation (DPA) <=> neural field

Distribution of population activation =
 $\sum_{\text{neurons}} \text{tuning curve} * \text{current firing rate}$

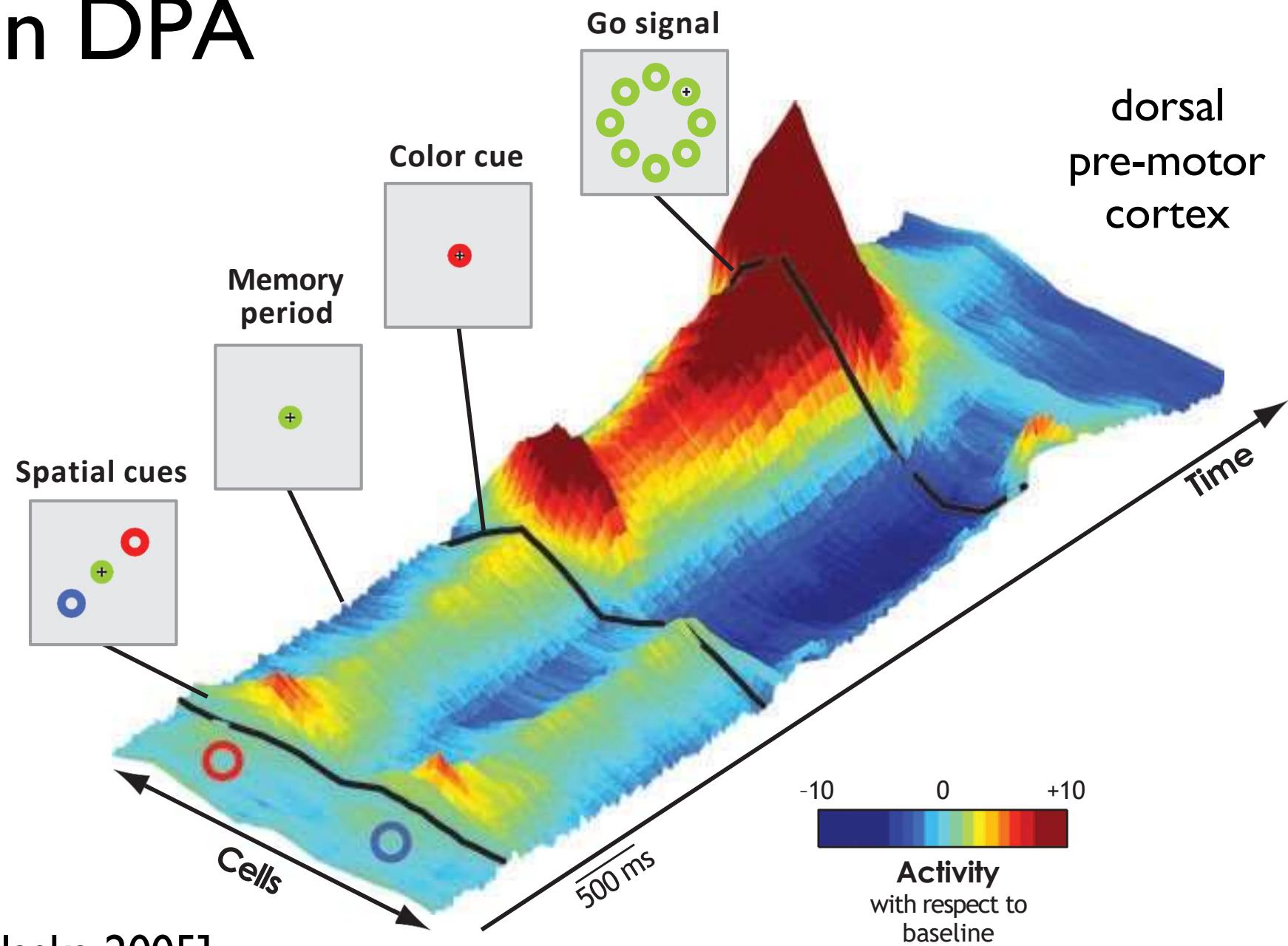


■ note: neurons are not localized within DPA!

[Bastian, Riehle, Schöner, 2003]

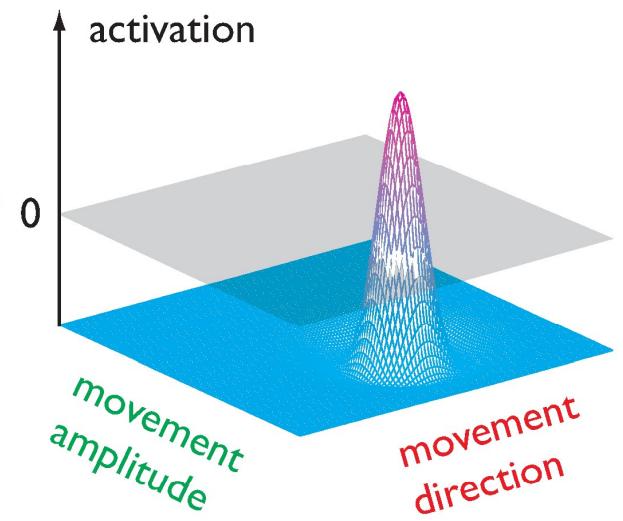
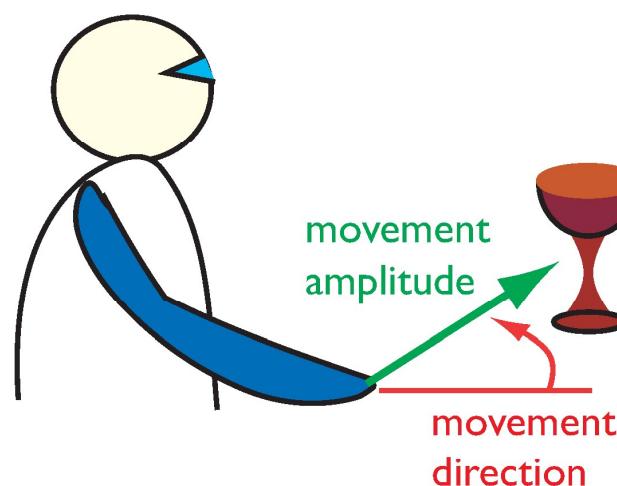
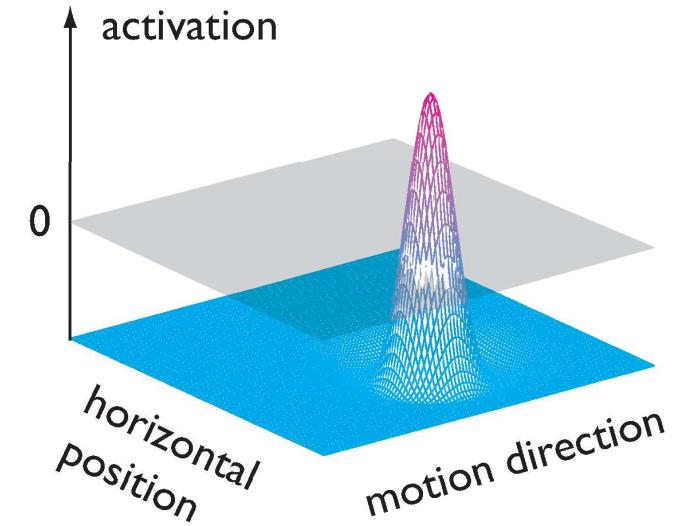
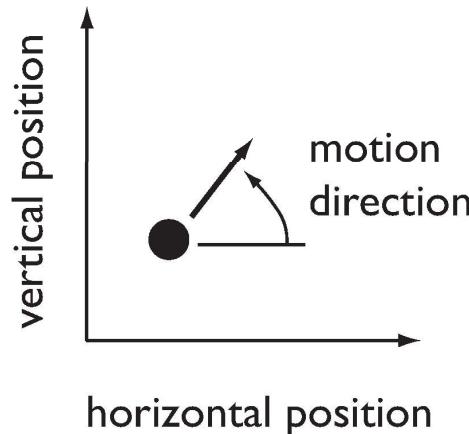
Decision making

in DPA



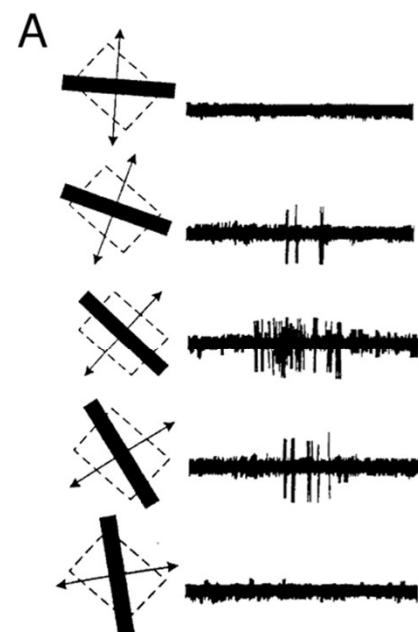
[Cisek, Kalaska 2005]

Fields may jointly represent different dimensions: examples



Neurons may be tuned to multiple different feature dimensions

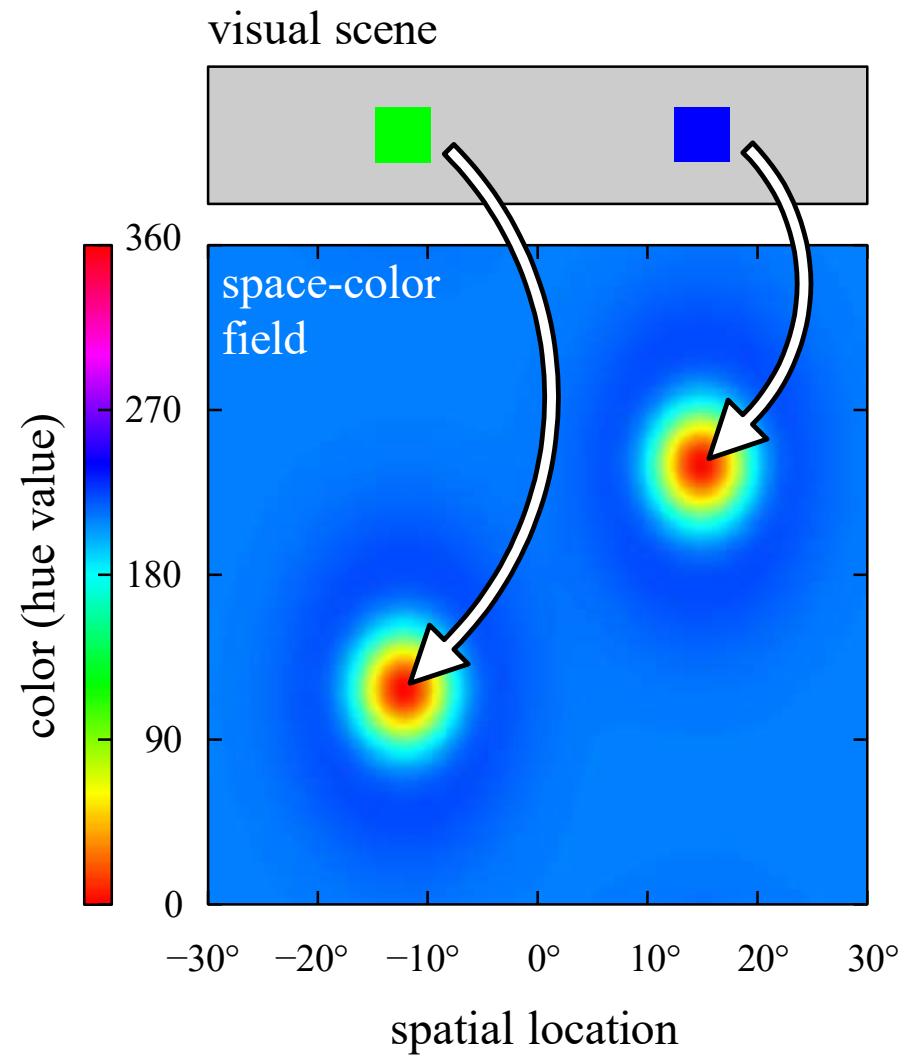
- example: receptive field + direction tuning
- => combines visual space and orientation
- => “anatomical” binding



[Hubel,Wiesel, 1962]

Combining different feature dimensions

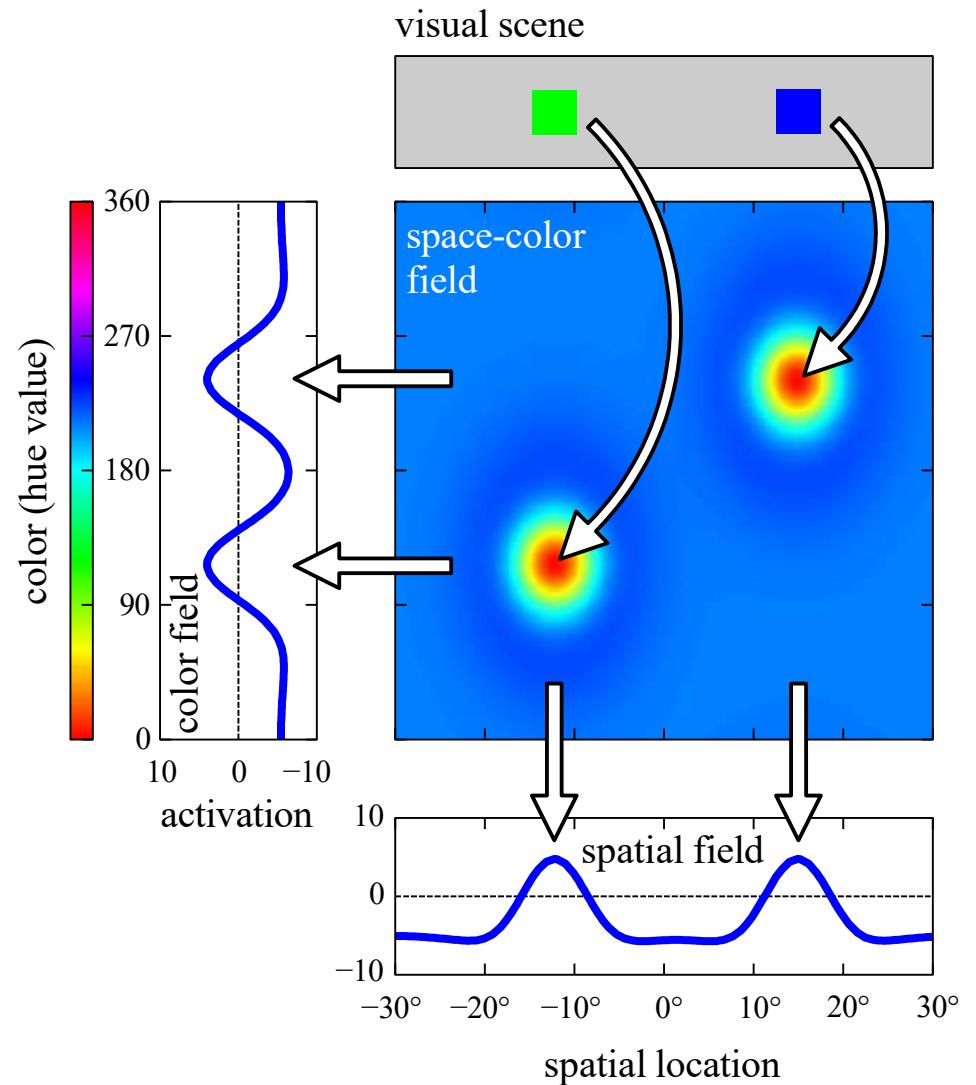
example: a joint representation of color and visual space “binds” these two dimensions



[Schneegans et al., Ch 5 of *DFT Primer*, 2016]

Extract the bound features

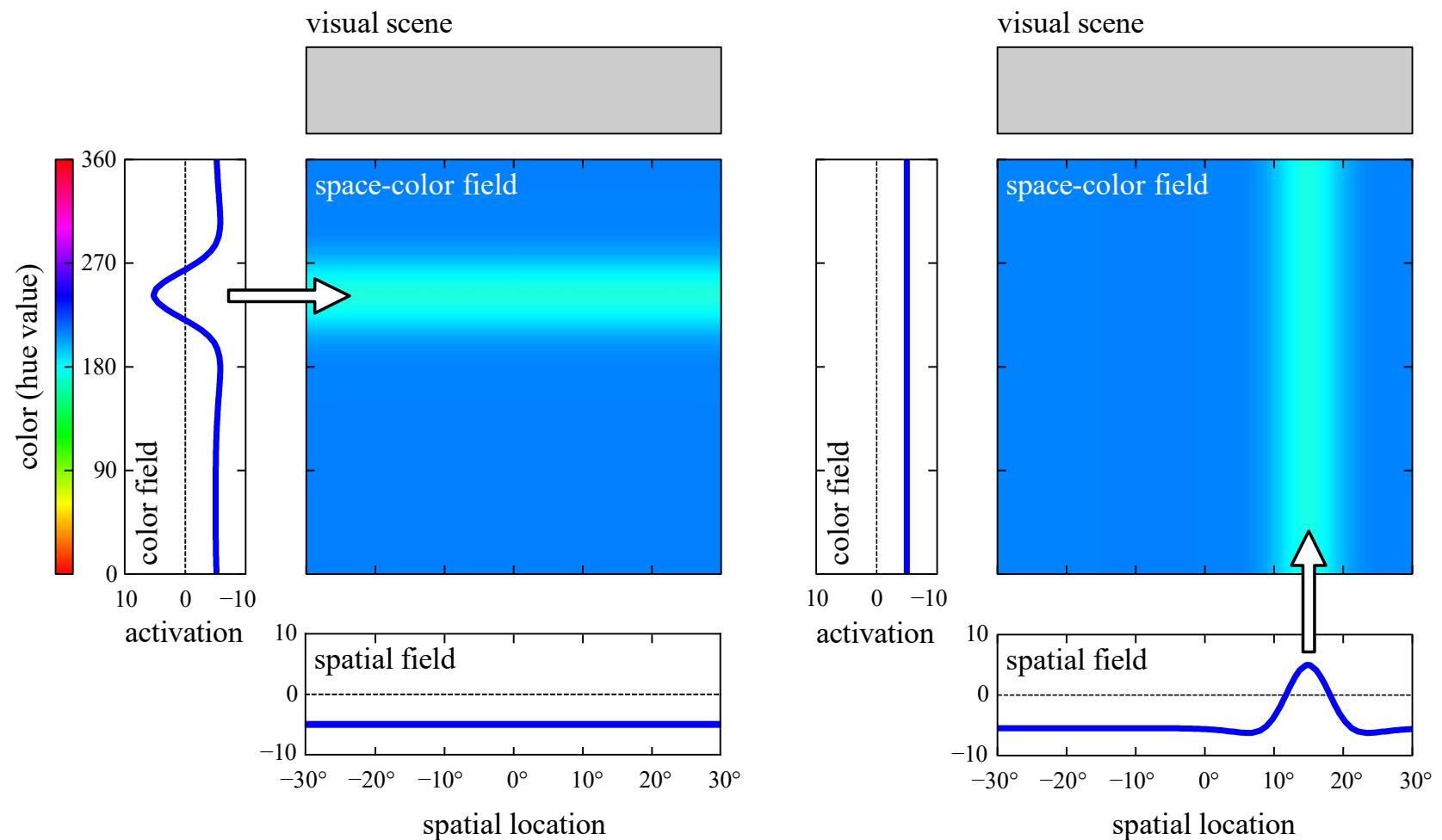
- project to lower-dimensional fields
- by summing along the marginalized dimensions
- (or by taking the softmax)



[Schneegans et al., Ch 5 of *DFT Primer*, 2016]

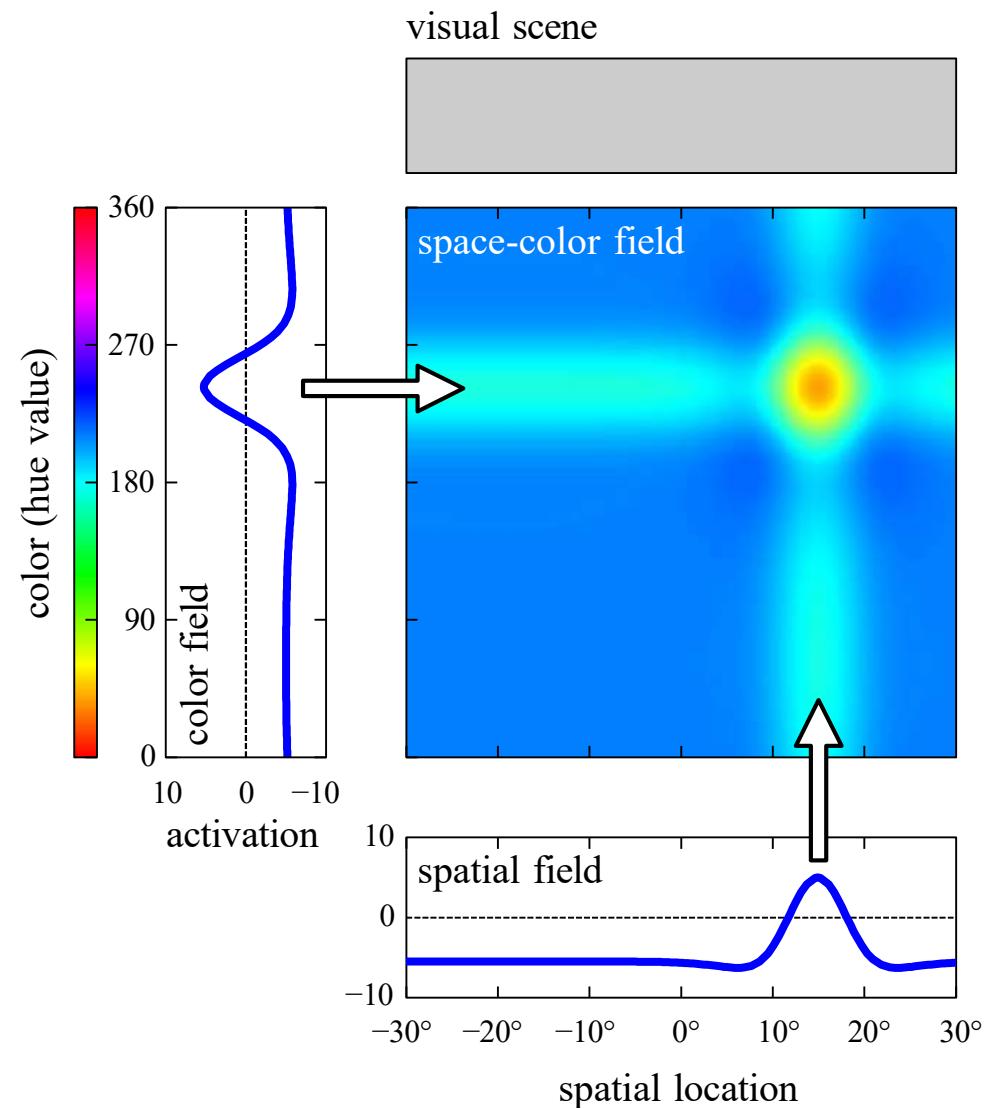
Assemble bound representations

project lower-dimension field onto higher-dimensional field as “ridge input”



[Schneegans et al., Ch 5 of DFT Primer, 2016]

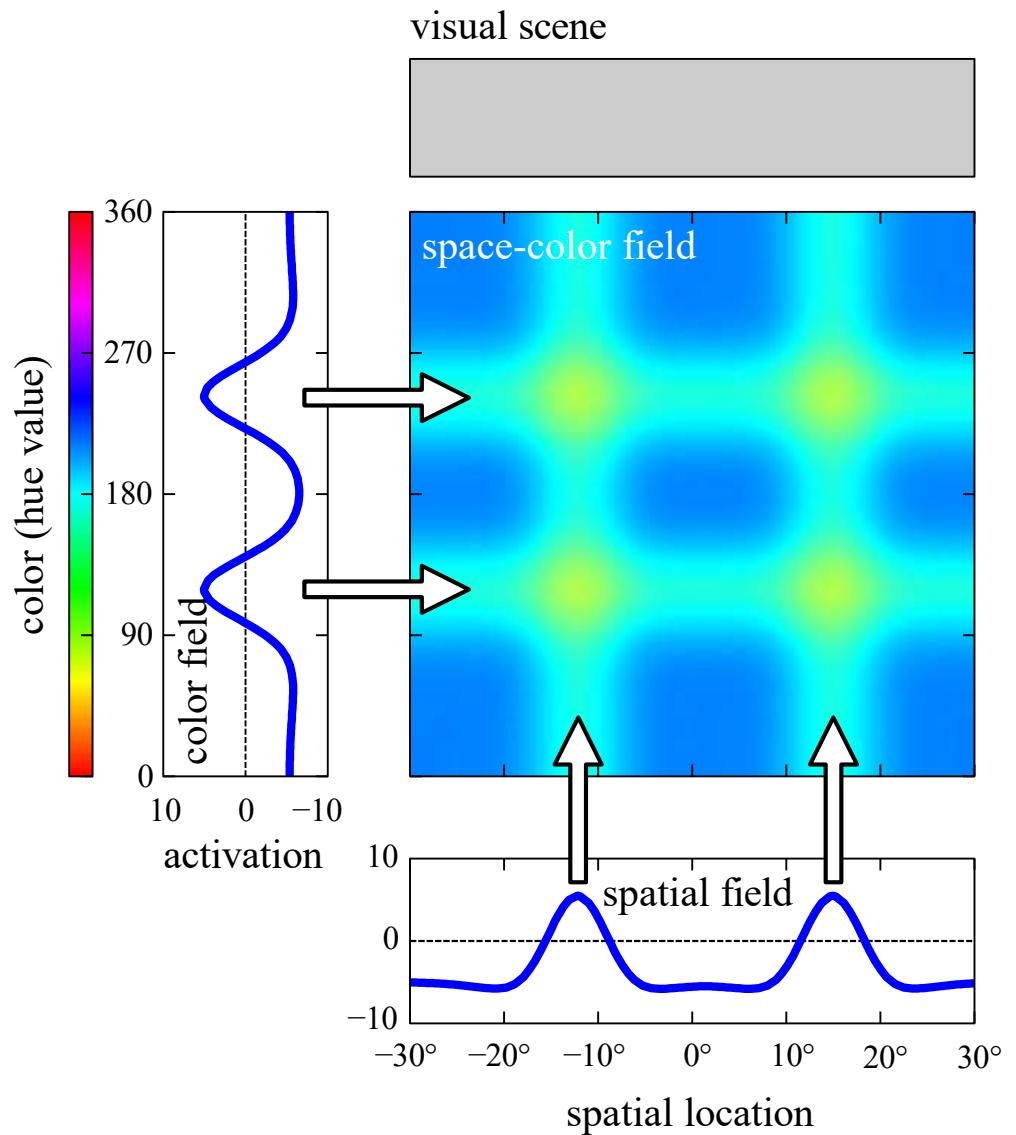
Assemble bound representations



[Schneegans et al., Ch 5 of *DFT Primer*, 2016]

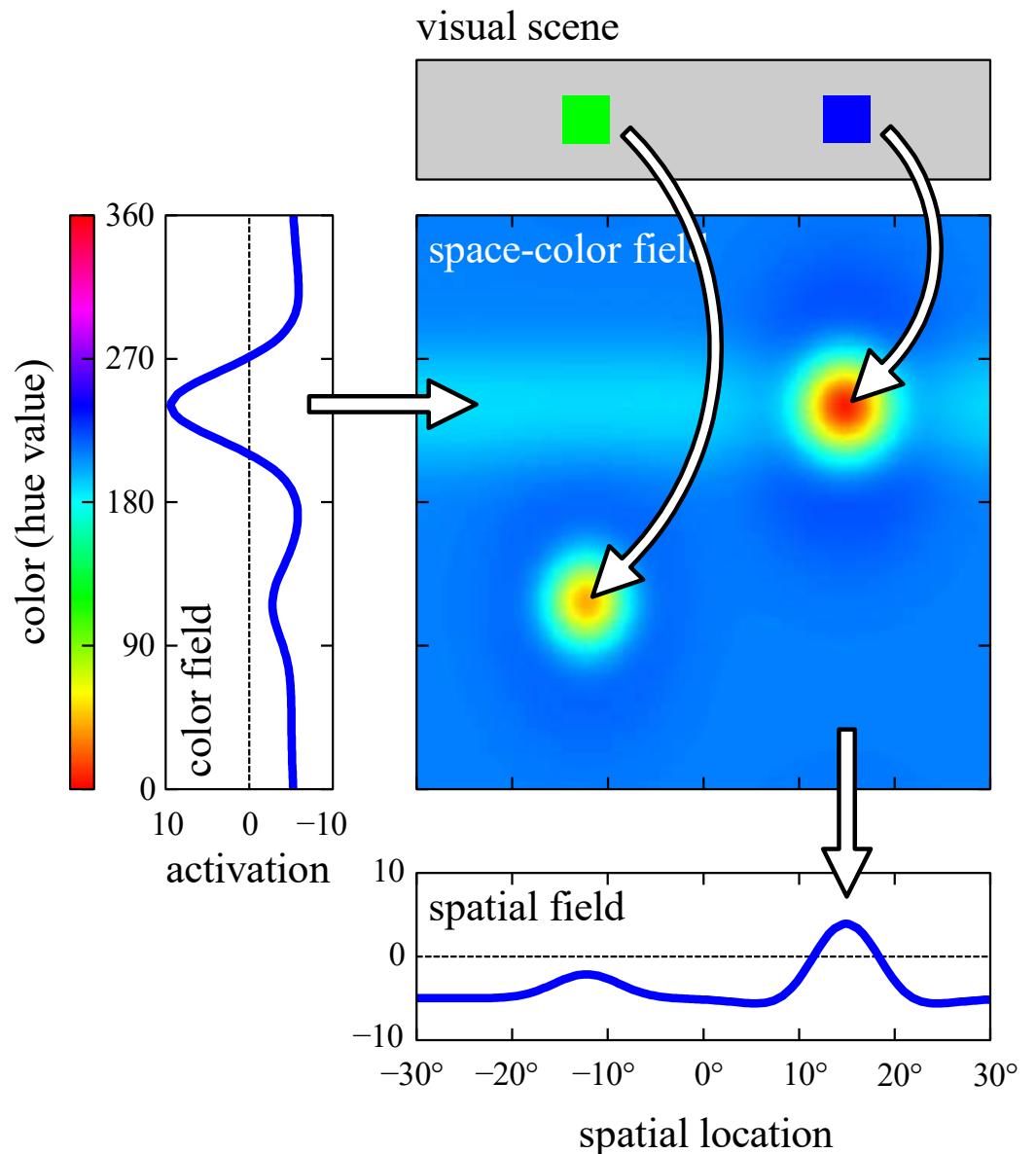
Assemble bound representations

- binding problem:
multiple ridges along
lower-dimensional
space lead to a
correspondence
problem
- => assemble one
object at a time...
- => sequentiality bottle-
neck!



Search

- ridge input along one dimension extracts from bound representation matching objects
- other dimensions of those objects can then be extracted
- e.g. visual search



[Schneegans et al., Ch 5 of *DFT Primer*, 2016]

Scaling feature dimensions

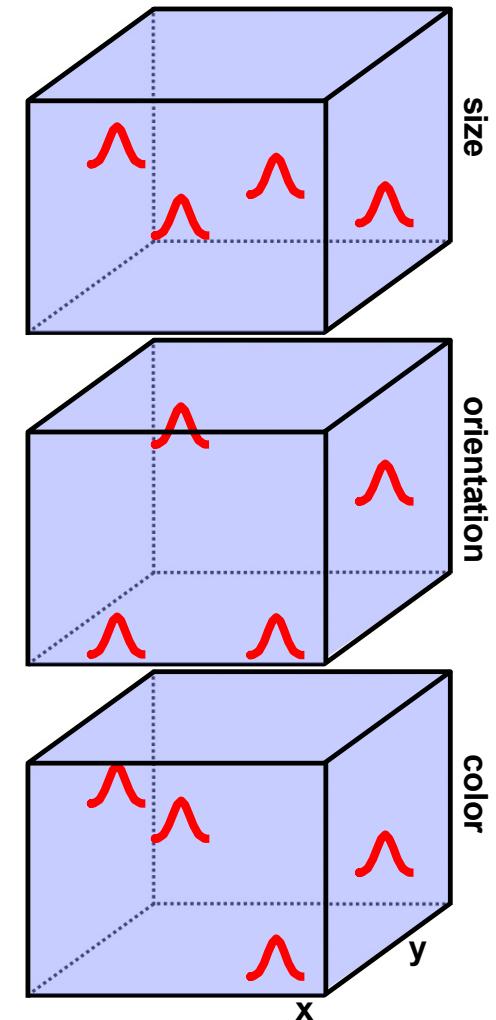
- 2 spatial dimensions
- depth
- orientation
- color
- texture
- movement direction
- size
- etc...

=>

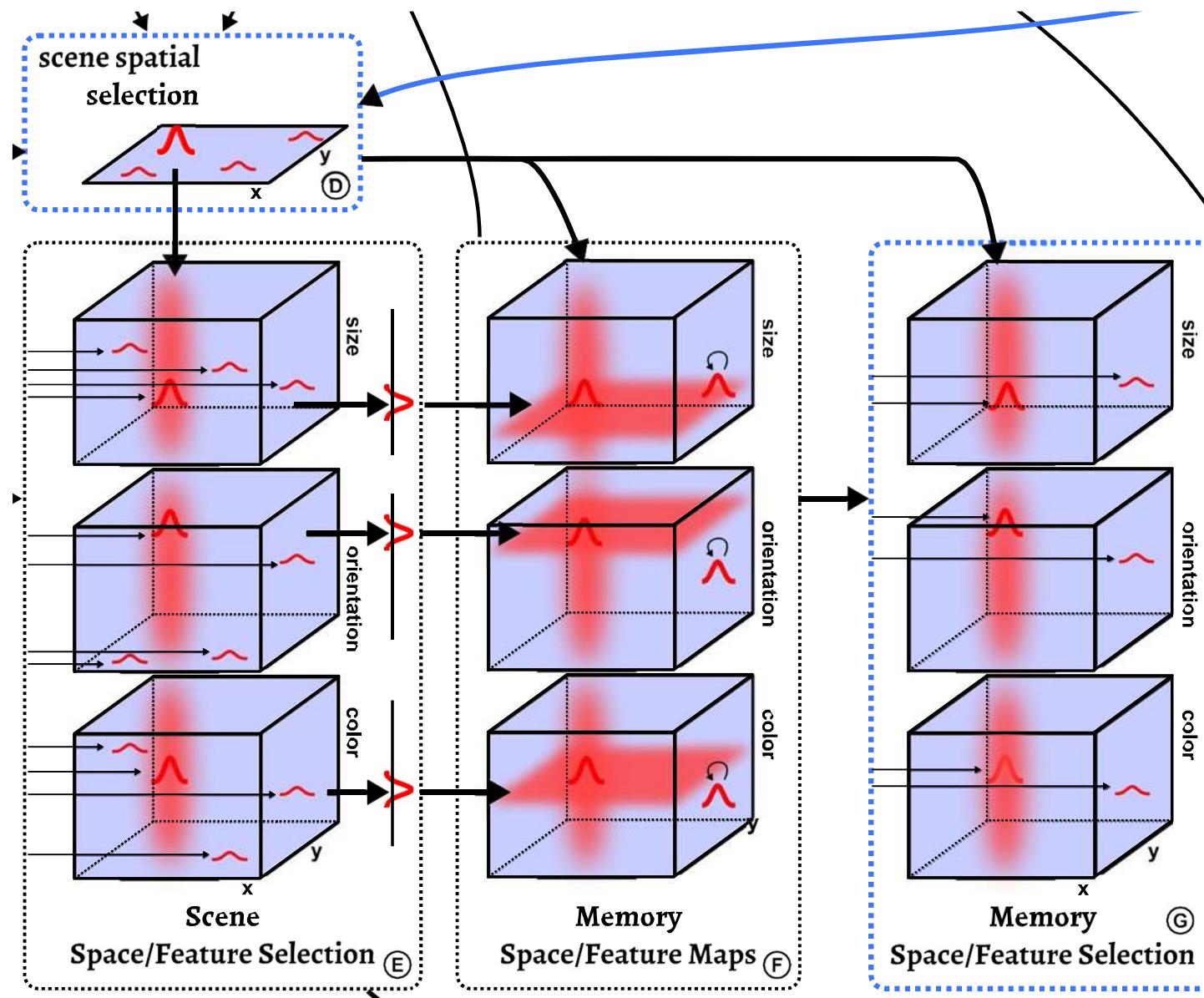
- e.g. 8 dimensions
- 100 neurons per dimension
- $10^{2*8} = 10^{16}!$
- more than there are in the entire brain!
- => only small sets of feature dimensions can be bound “anatomically”

Binding through space

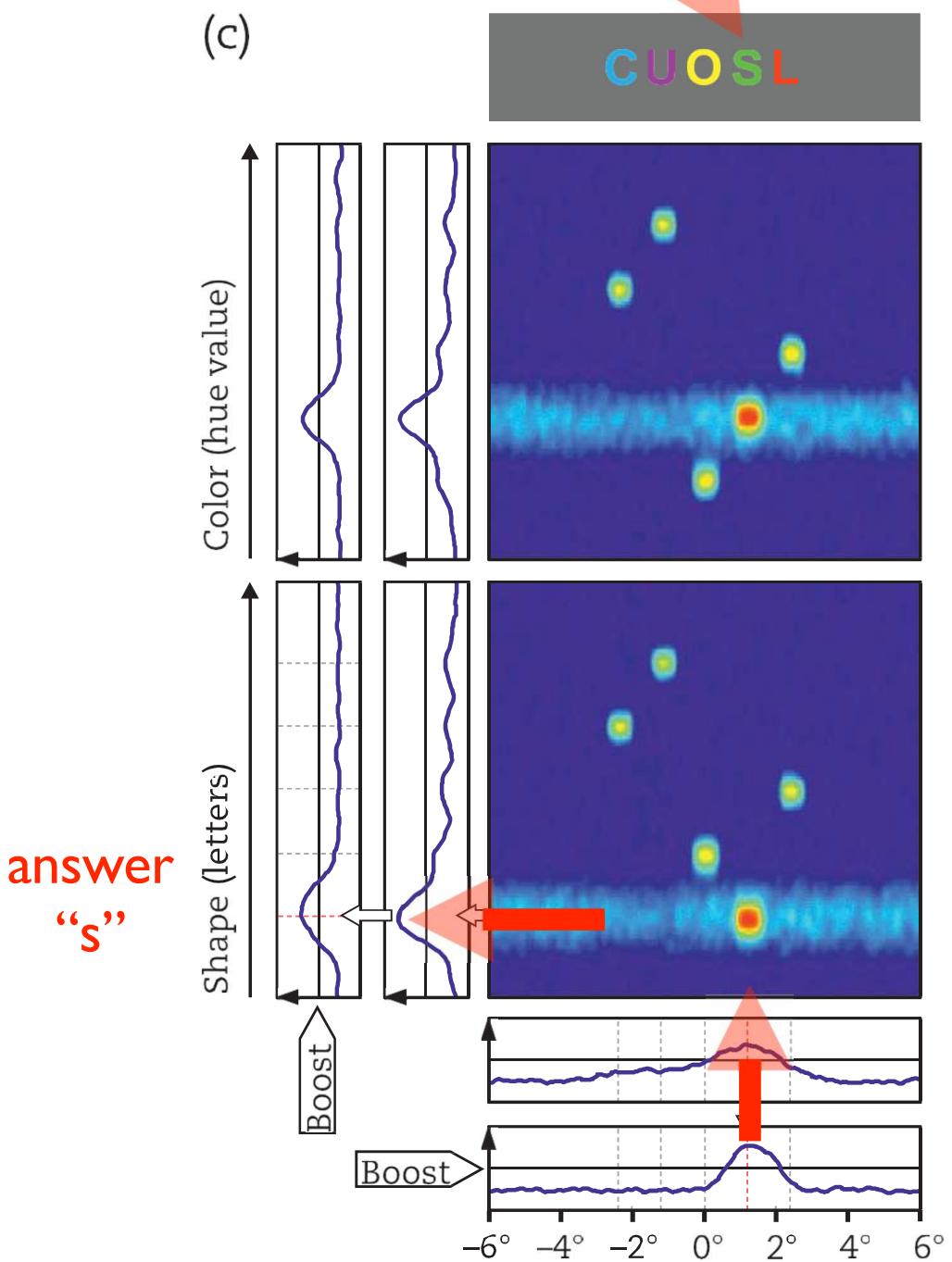
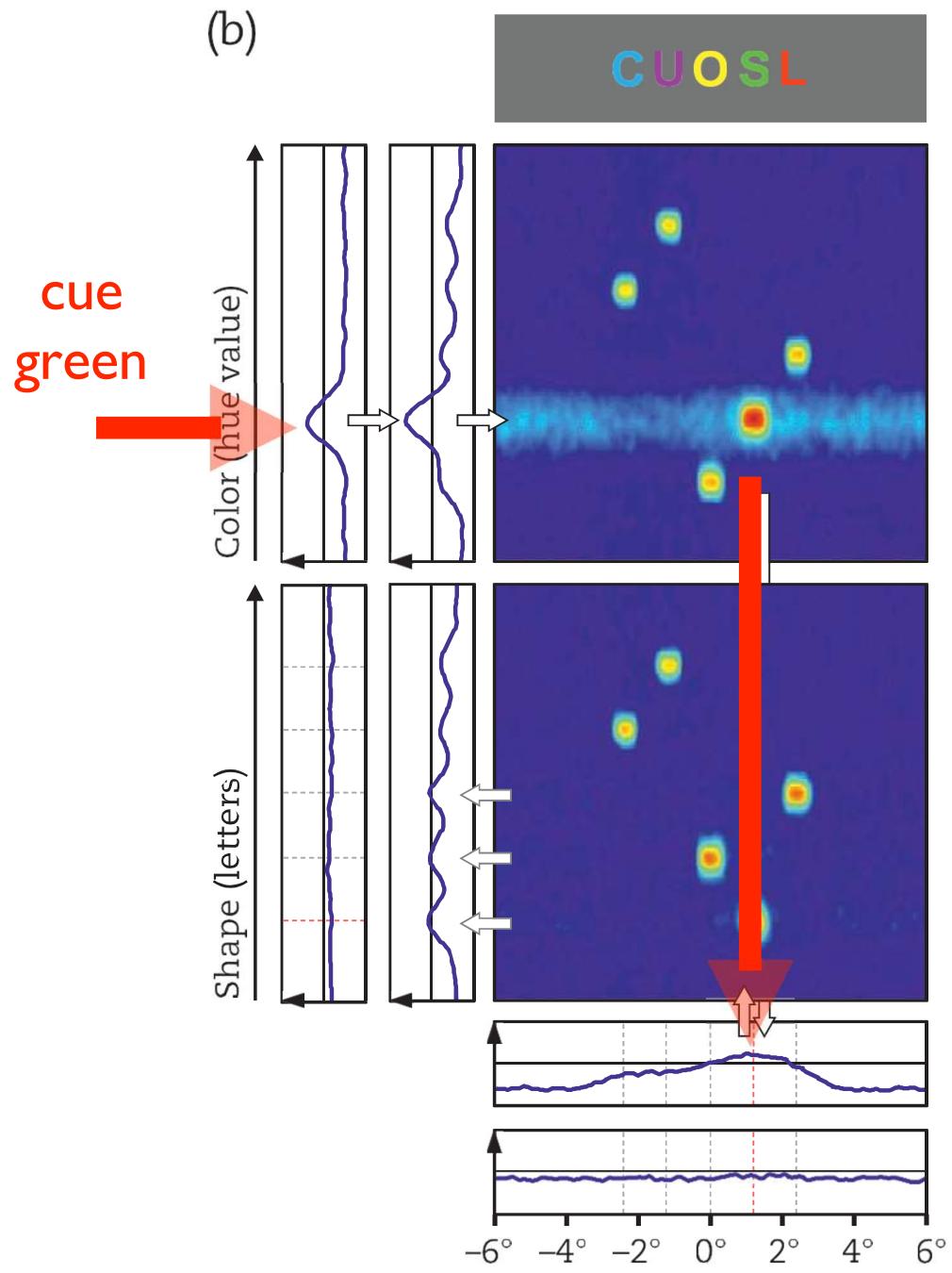
- many 3 to 4 dimensional feature fields
- all of which share the one dimension: visual space (~all neurons have receptive fields)
- bind through space à la Feature Integration Theory (Treisman)



Binding through space

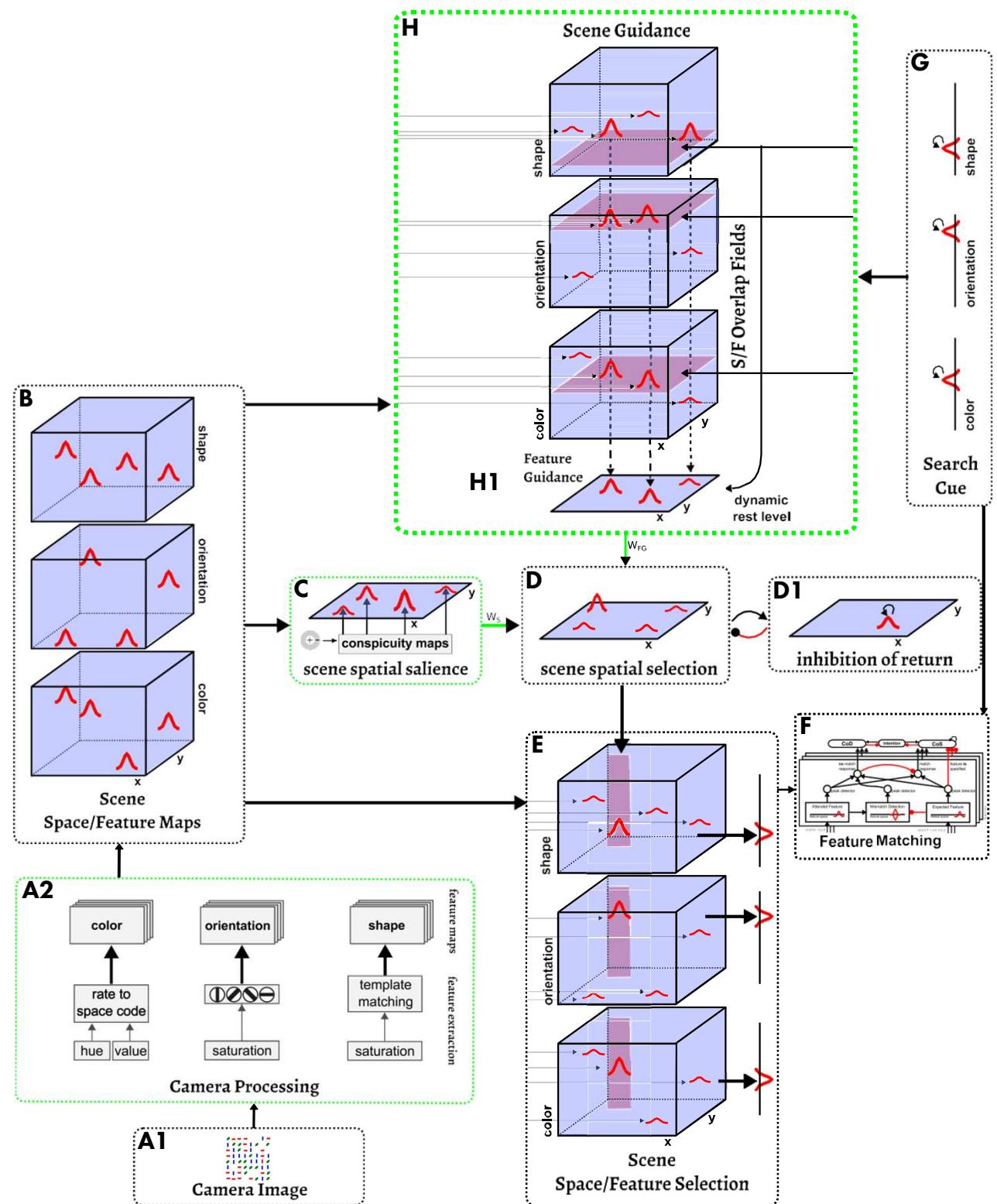


[Grieben et al. *Attention, Perception & Psychophysics* 2020]



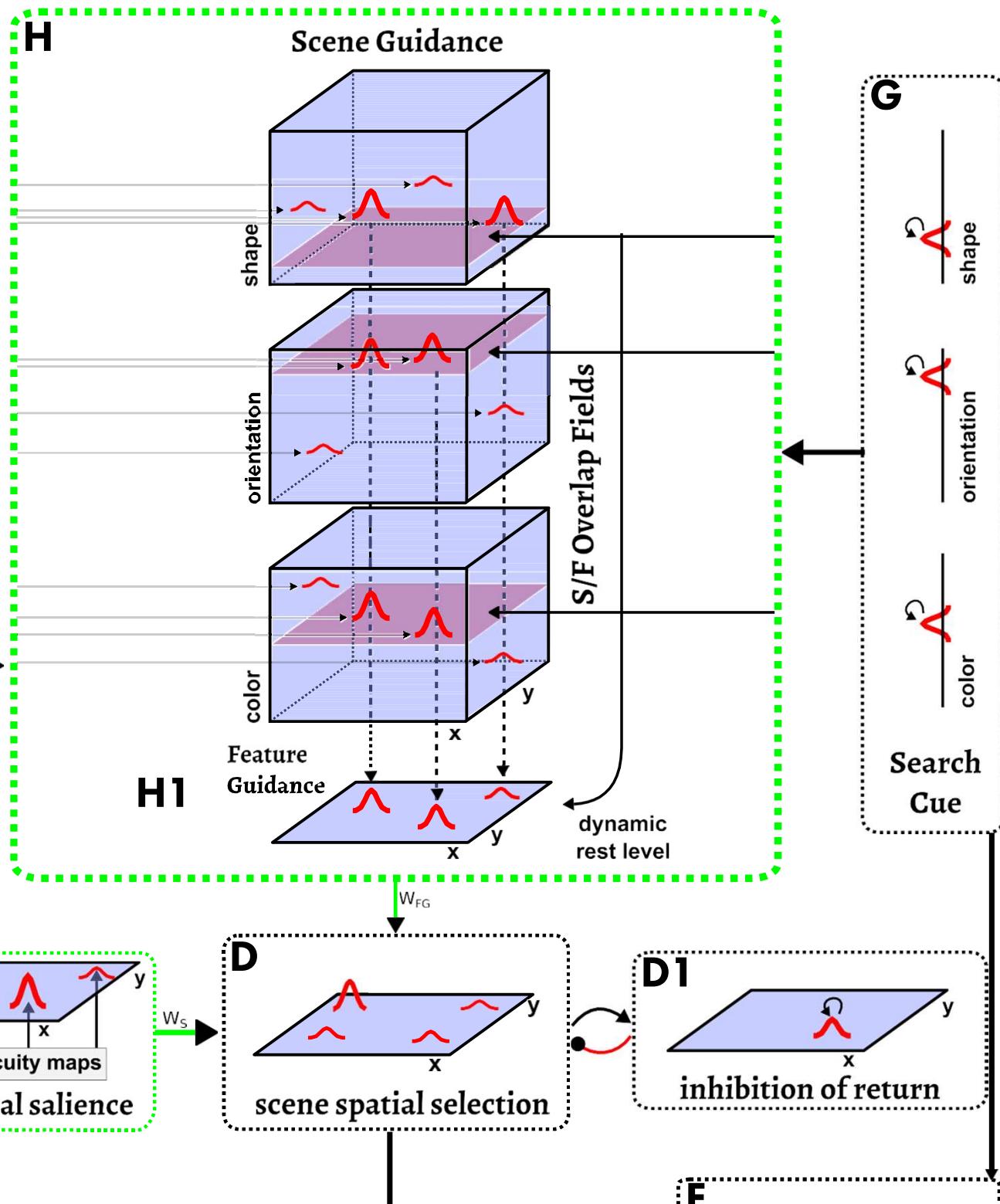
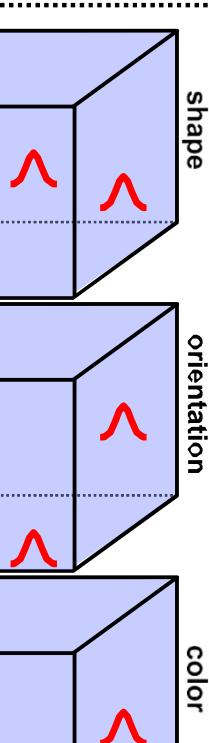
[Schneegans et al., Ch 5 of DFT Primer, 2016]

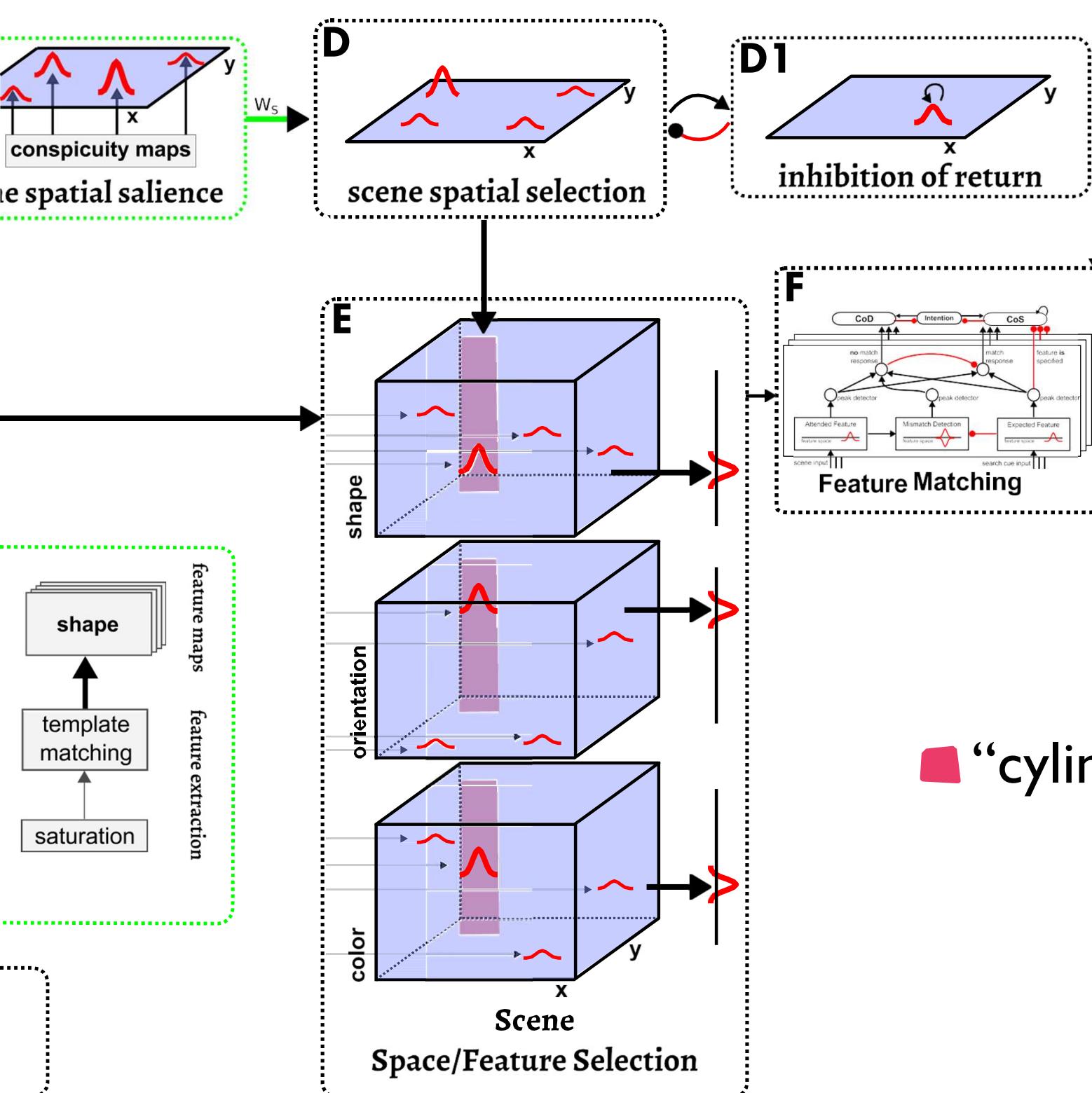
Conjunctive visual search



[Griebel et al. *Attention, Perception & Psychophysics*, 2020; CogSci 2021]

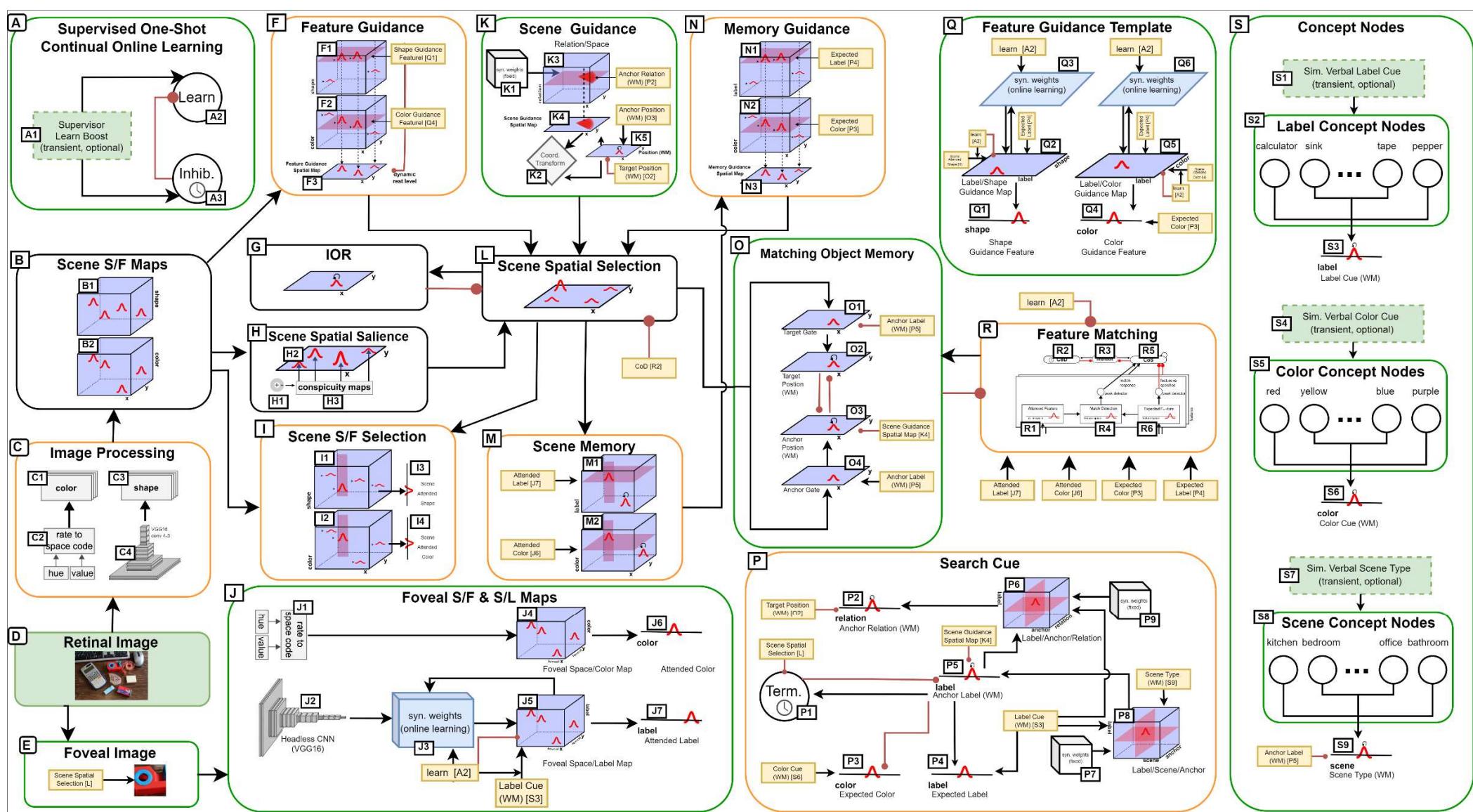
“slice” input





Visual search







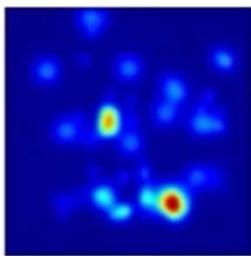
Camera Image



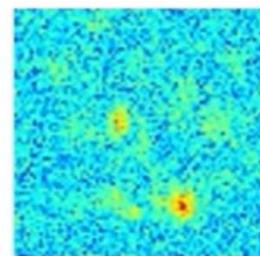
Foveal Image



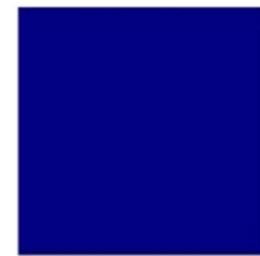
Target Position (WM)



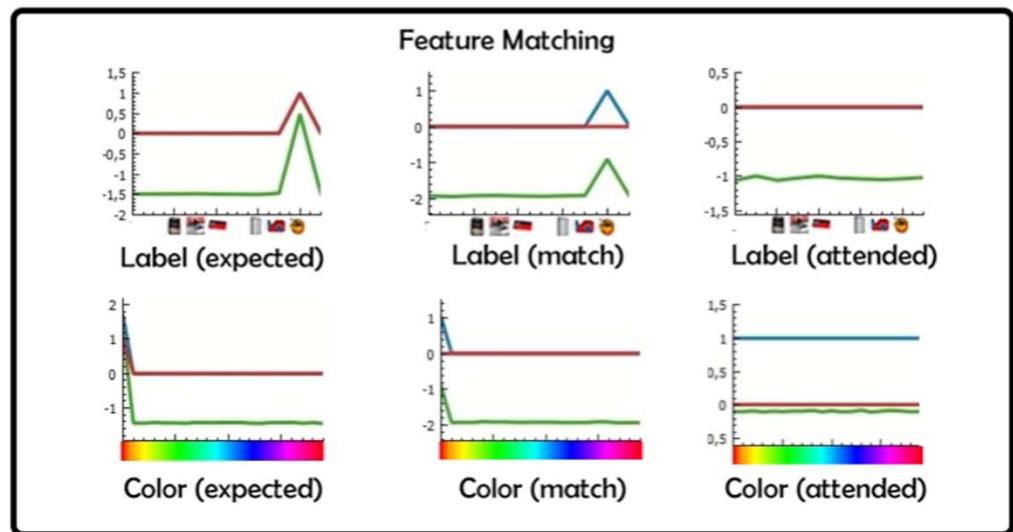
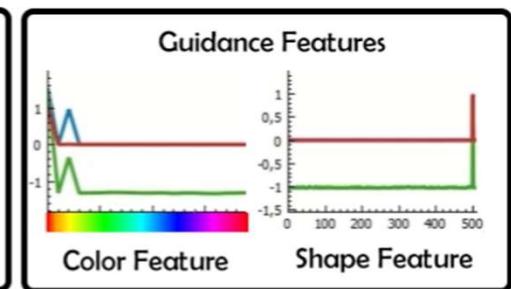
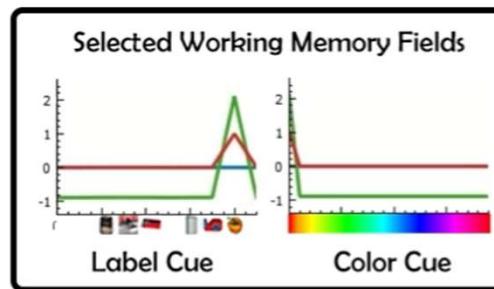
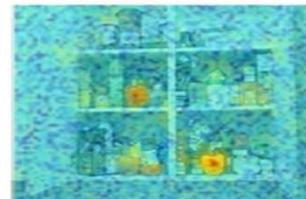
Attention (Input)



Attention (Activation)

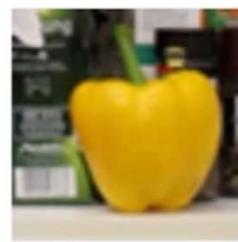


Attention (Sig. Activation)

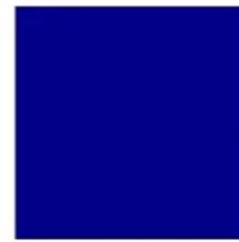




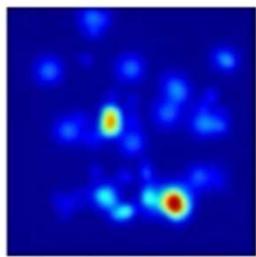
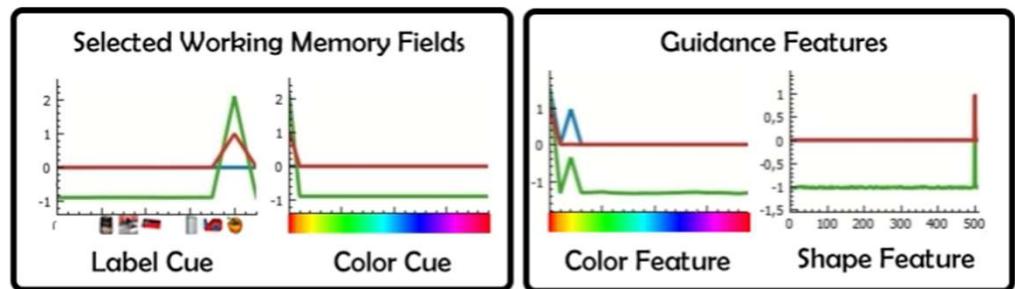
Camera Image



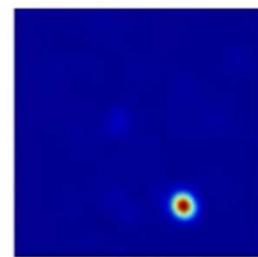
Foveal Image



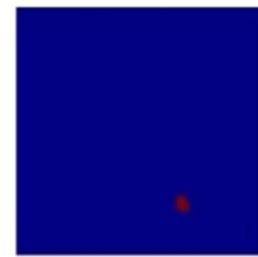
Target Position (WM)



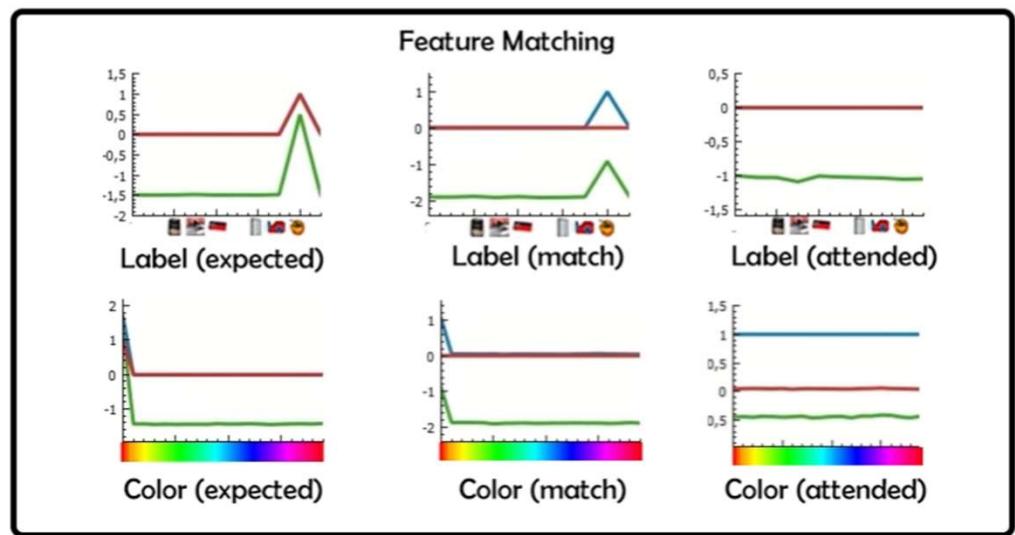
Attention (Input)



Attention (Activation)



Attention (Sig. Activation)





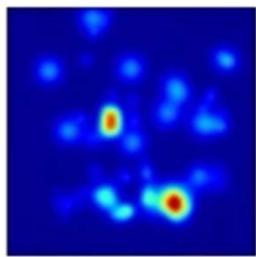
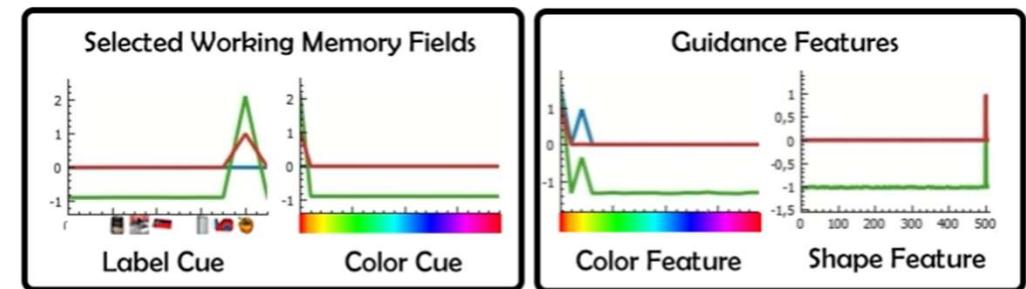
Camera Image



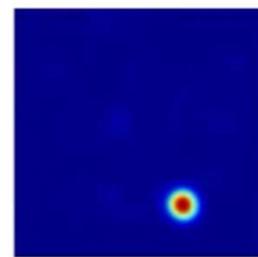
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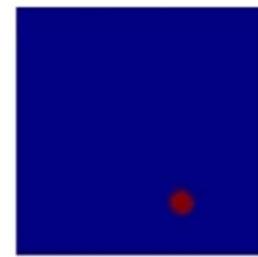
Target Position (WM)



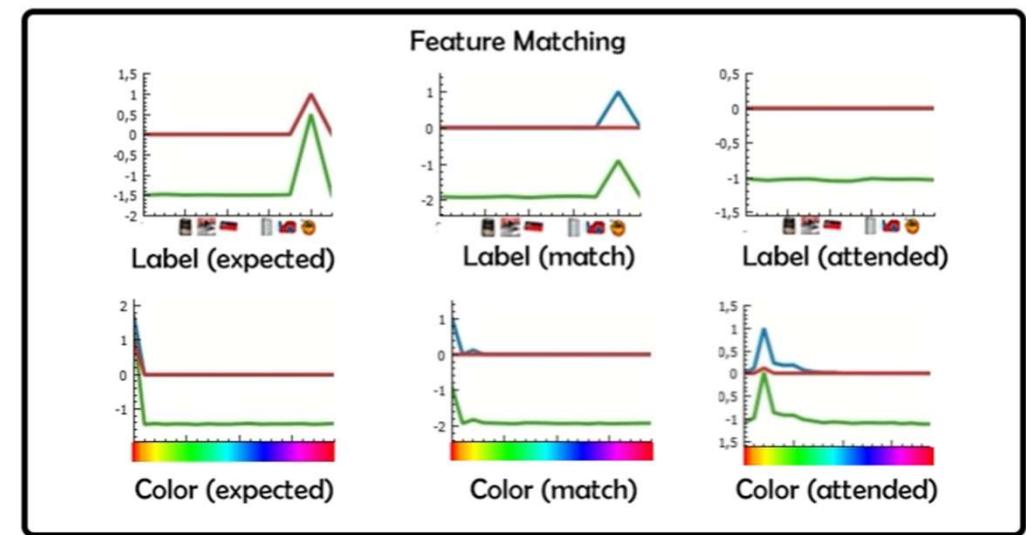
Attention (Input)



Attention (Activation)

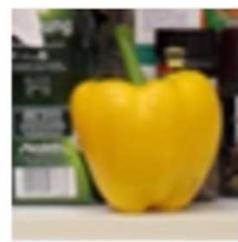


Attention (Sig. Activation)





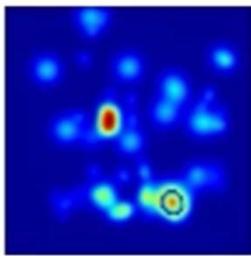
Camera Image



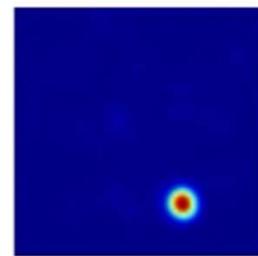
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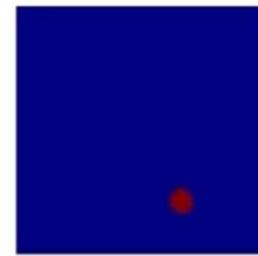
Target Position (WM)



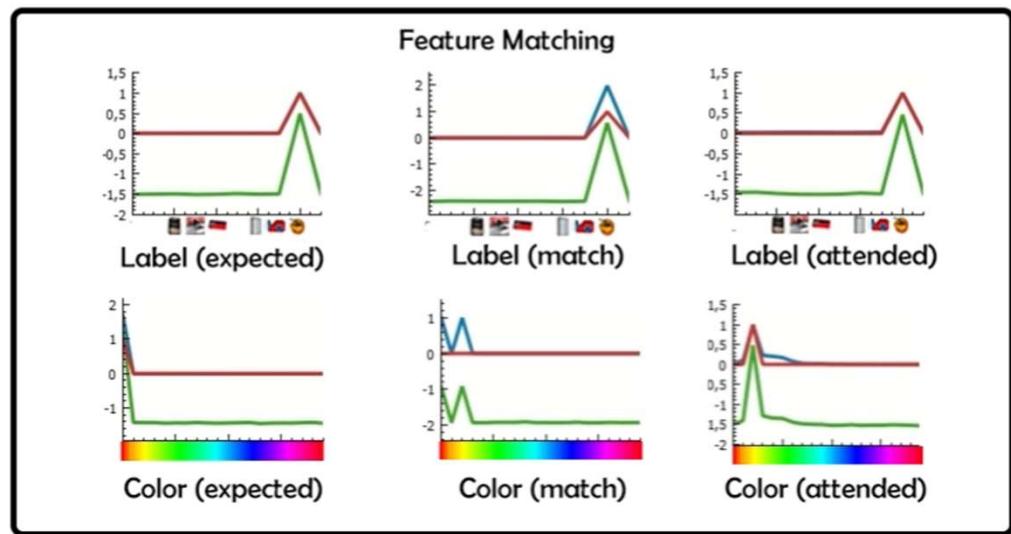
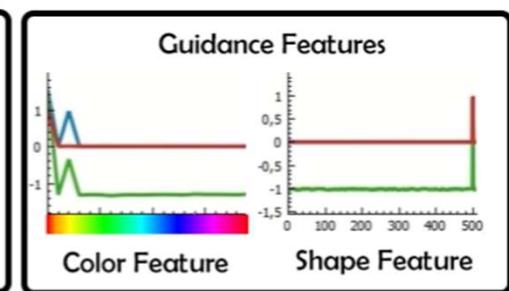
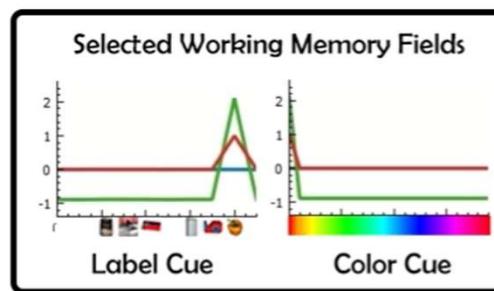
Attention (Input)



Attention (Activation)



Attention (Sig. Activation)





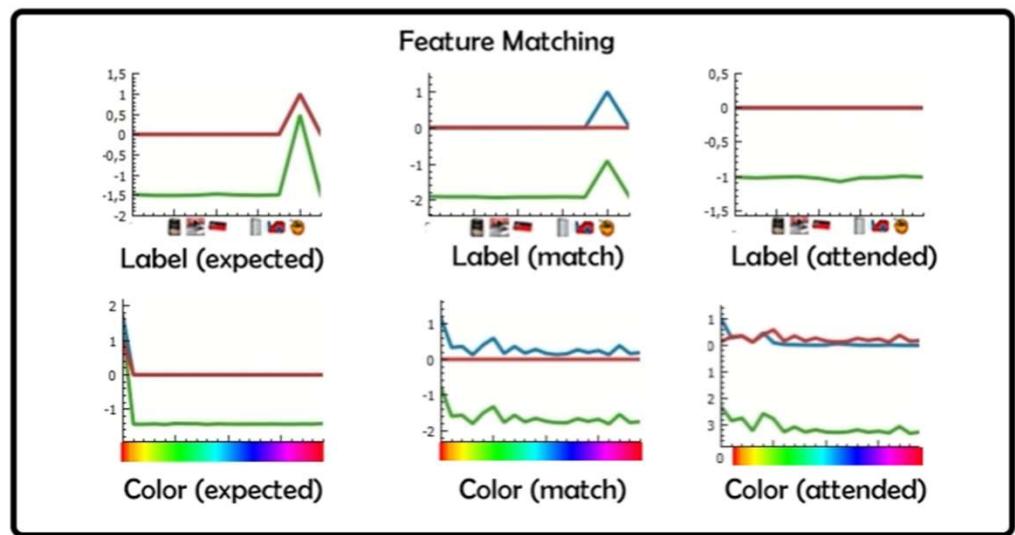
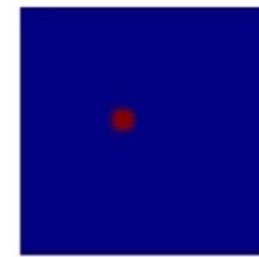
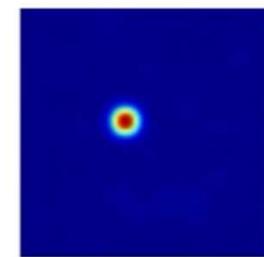
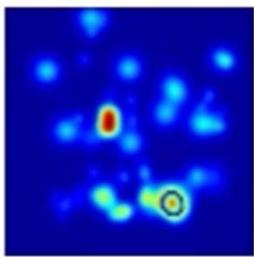
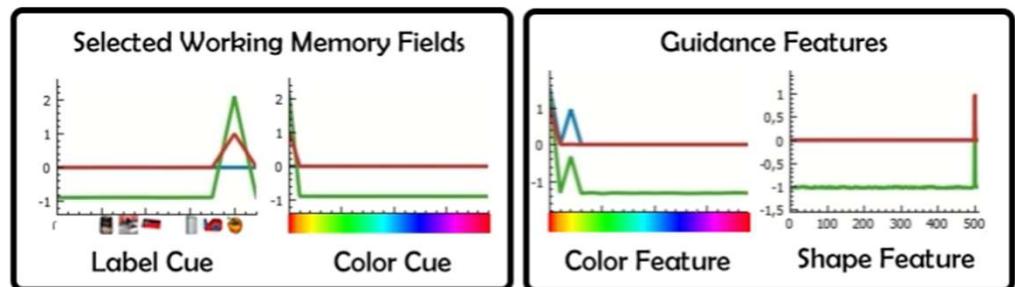
Camera Image



Foveal Image



Target Position (WM)





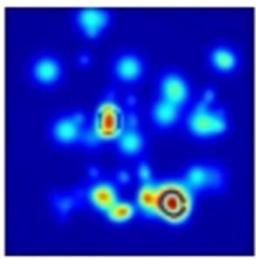
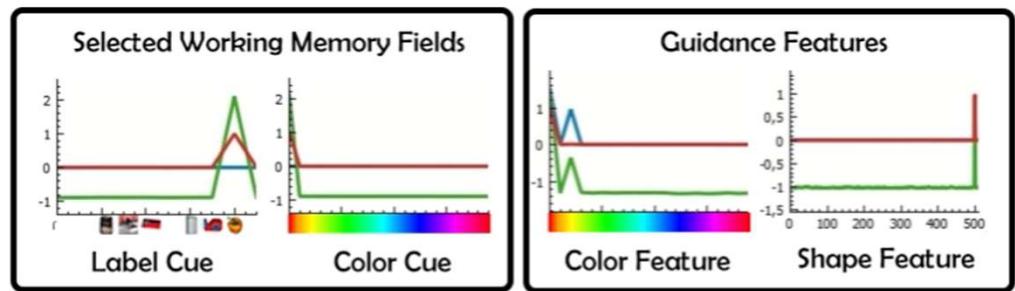
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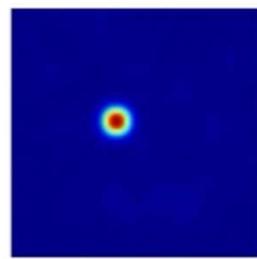
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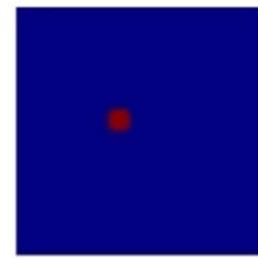
Target Position (WM)



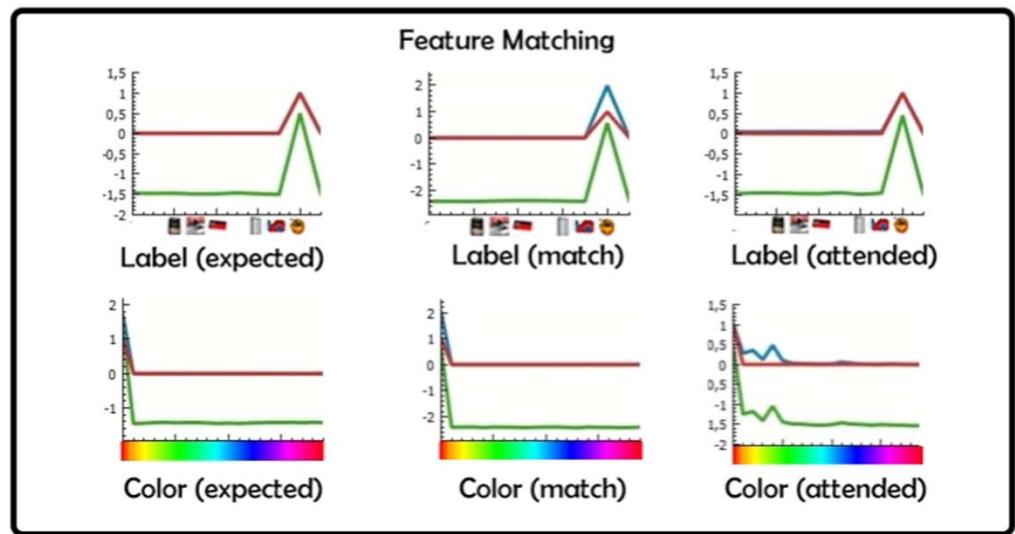
Attention (Input)



Attention (Activation)



Attention (Sig. Activation)



- Neural fields: dimensions
- Binding
- Visual search