

# Spatial context, salience and eye movements



Computational neuroscience class  
Odelia Schwartz, 2020

- Start with zoom introductions ...
- Mute microphone unless asking questions
- Turn video off if too slow
- Give me feedback! (email / in class)

- Matlab on your computers
- Final Projects / assignments
- Other questions

# Matlab on your computer

Visit: <https://www.it.miami.edu/a-z-listing/matlab/index.html>

MATLAB - University of Miami

Description. MATLAB® is a high-level language and interactive environment for numerical computation, visualization, and programming. Using MATLAB, you can analyze data, develop algorithms, and create models and applications.

[www.it.miami.edu](http://www.it.miami.edu)

Select scroll to the bottom for downloading instructions

Email: [software@miami.edu](mailto:software@miami.edu) (to request access to the UM's license)

UMIT will reply back with an activation key and installation key.

You can then download and use Matlab on your personal computer.

Please note that in the lab we have been using R2017b. I have not checked compatibility of later versions of Matlab.

Here is what I downloaded to my home laptop, though we are not actually using all of the noted toolboxes:

Matlab

(Simulink: we are not using)

Audio

Curve fitting

Image processing

Neural Network

Optimization

Signal processing

Statistics and machine learning

Symbolic math

DSP systems toolbox



# Final projects / assignments

As per the last email:

- Since Spring break has been extended by one week, you can submit the current assignment up to one week past the due date (so now due March 26).

- I would like to make the final projects more flexible, choosing one of the following options:

1. In lieu of group projects, you can hand in an individual assignment, in which you choose two of the upcoming labs, explain the labs and answer the questions at the end.

2. If you already have a group that you would like to work with on a project (and are comfortable doing so remotely), or if you are keen on doing an individual project, then please send me an email. Projects can be done in any programming language. Please email me.

3. Students who do not have a CS or Engineering background can hand in a discussion about a computational neuroscience paper. You can choose one of the papers that we discuss in class (I will make this more explicit, providing a choice of papers).

# Spatial context



# Spatial context



# Contextual influences

- Perceptual illusions: “no man is an island..”



Review paper on context:  
Schwartz, Hsu, Dayan, Nature Reviews Neuroscience 2007

# Contextual influences

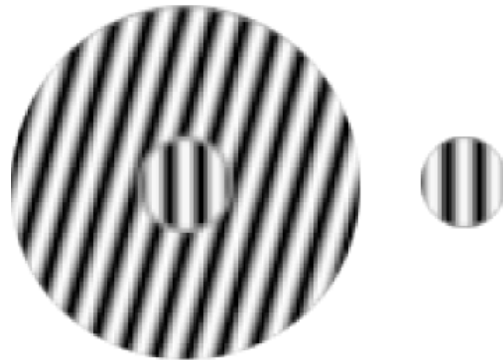
- Perceptual illusions: “no man is an island..”



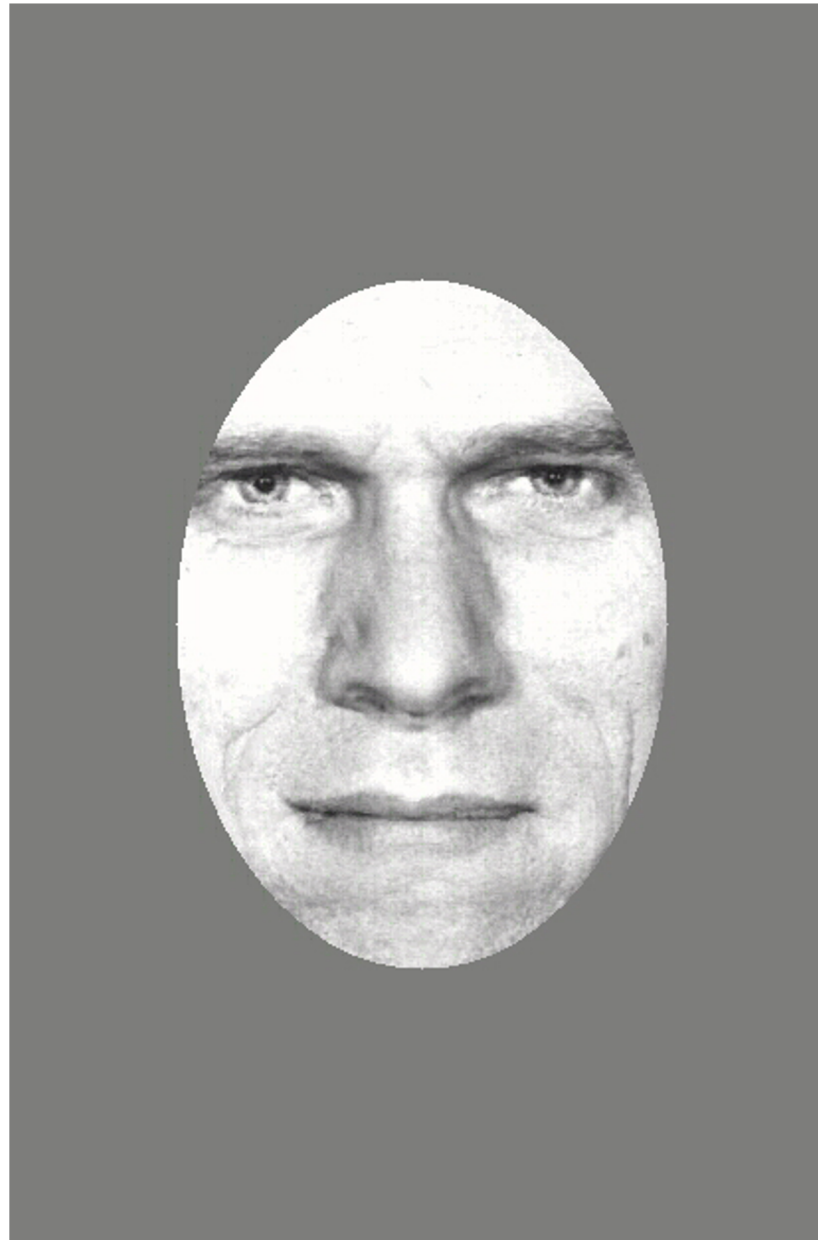
Review paper on context:  
Schwartz, Hsu, Dayan, Nature Reviews Neuroscience 2007

# Contextual influences

- Perceptual illusions



Contextual effects in time...



Adaptation to expression: pre-adapt (from Michael Webster)



adapt

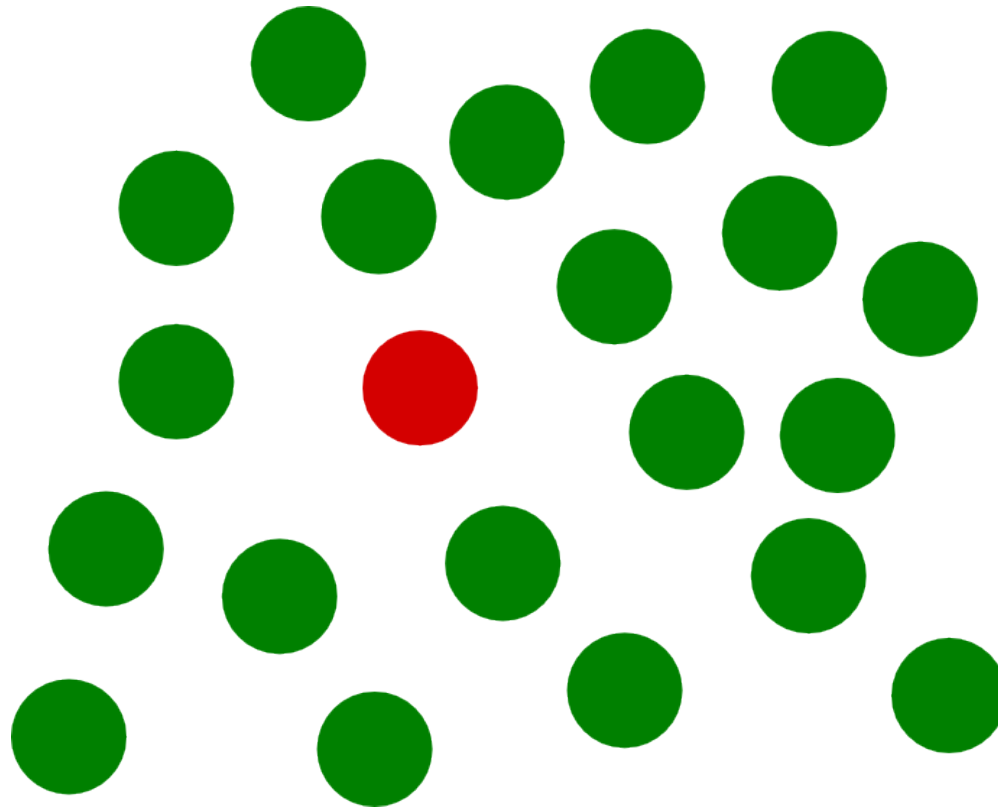




post-adapt

# Contextual influences

- Visual salience

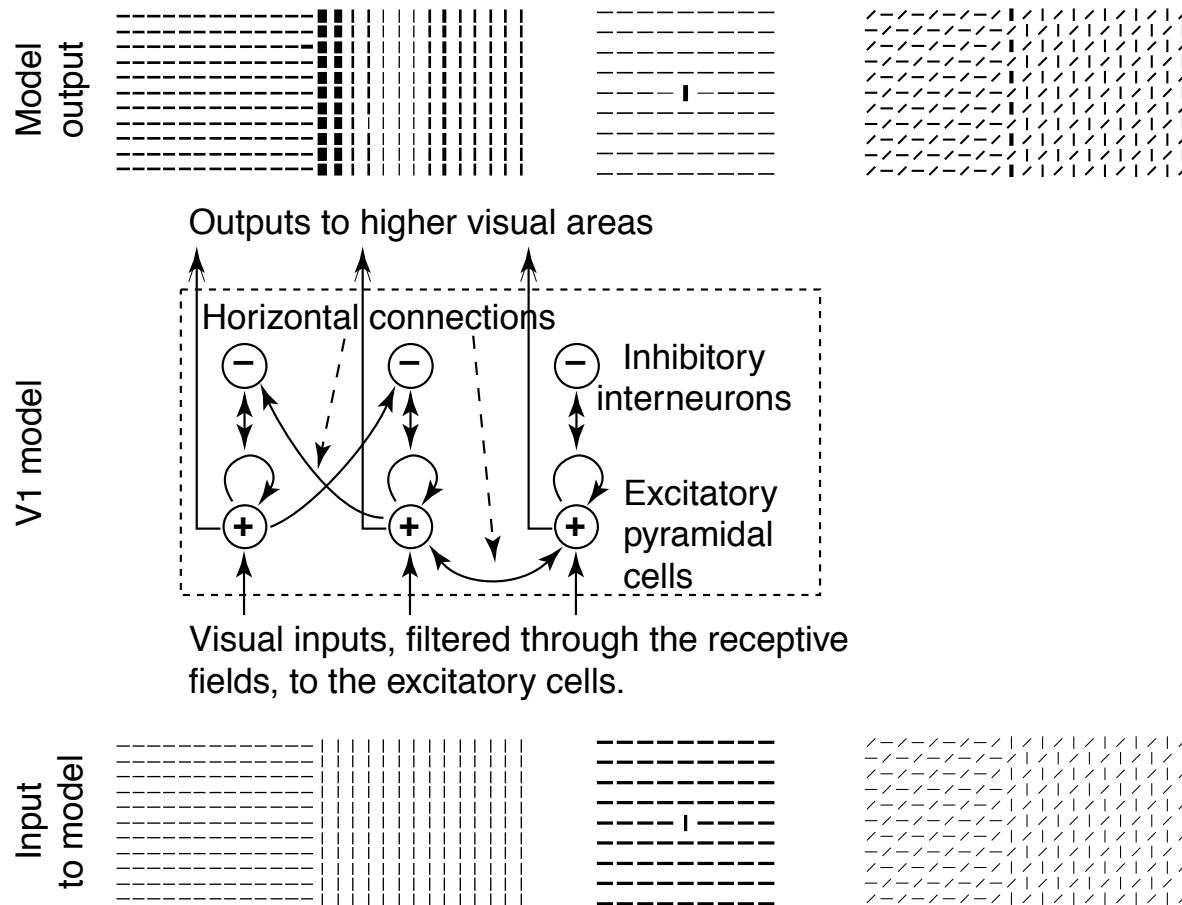


# Contextual influences

- Visual salience



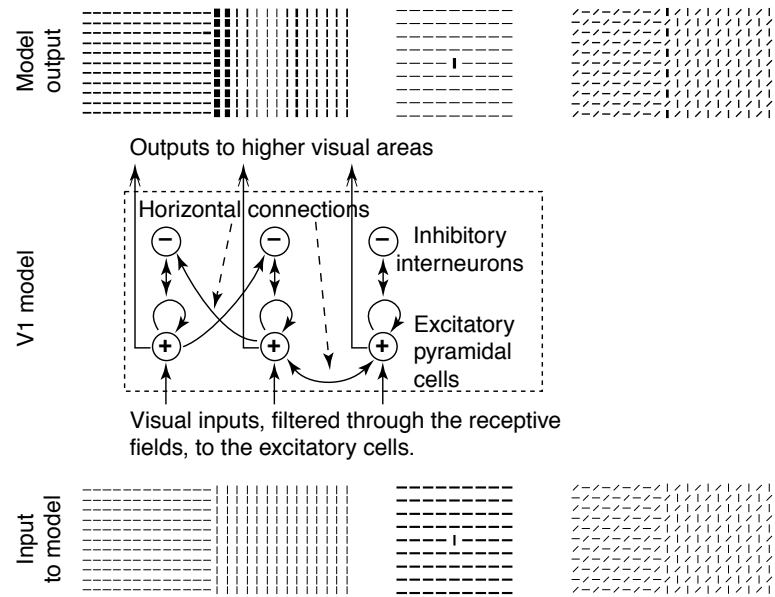
# Saliency model of V1 (Zhaoping)



*TRENDS in Cognitive Sciences*

Li Zhaoping, Trends in Cognitive Sciences, 2002.

# Saliency model of V1 (Zhaoping)



*TRENDS in Cognitive Sciences*

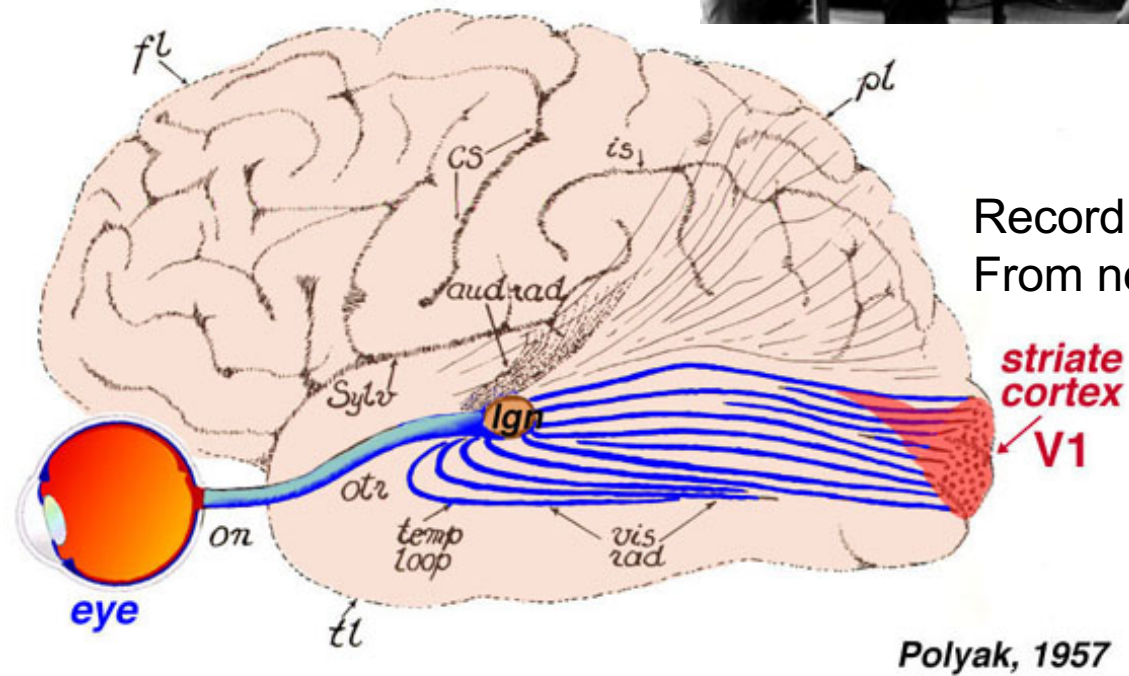
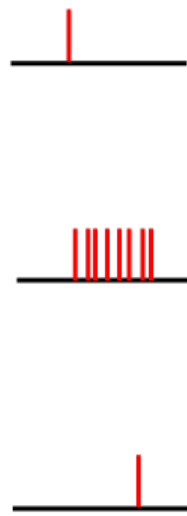
- Dynamical circuit model
- V1 saliency map
- Saliency as breakdown of statistical homogeneity

Li Zhaoping, Trends in Cognitive Sciences, 2002.

**Surround context (non classical  
receptive field) effects in visual  
neurons**

# What about neurons?

- Cortical neural processing

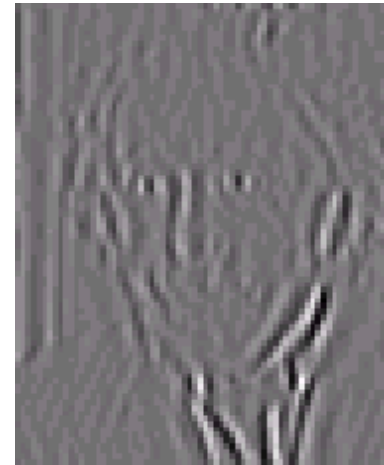
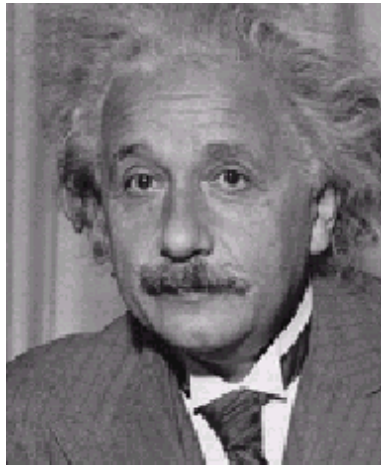


Record  
From neuron

Polyak, 1957

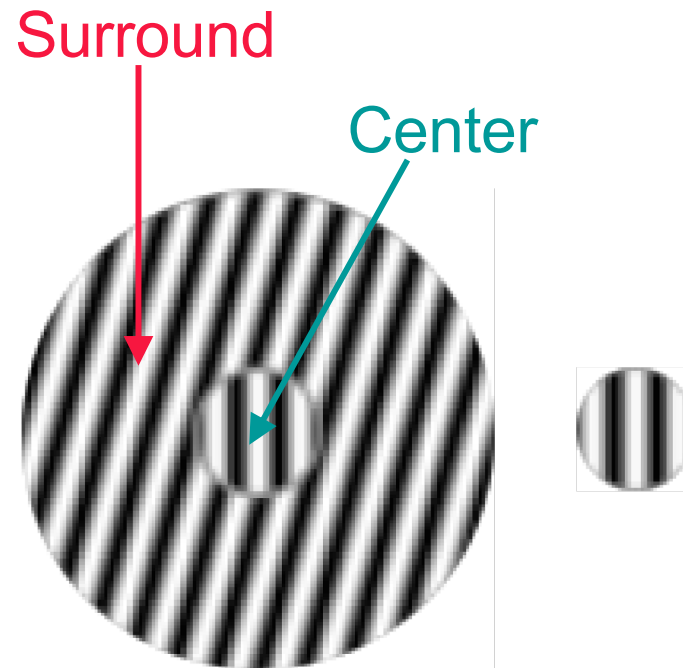
# What about neurons?

- Computer science / Engineering:  
visual receptive field or filter



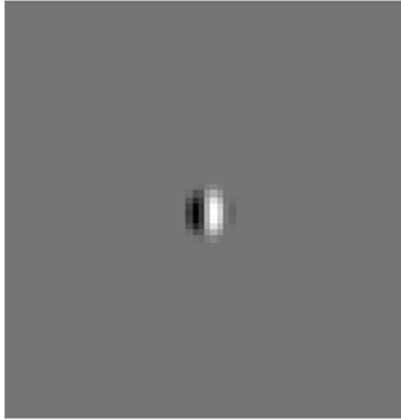


# Focus: spatial surround context



# Visual cortex: non classical RF

Center  
(classical RF)



Large response

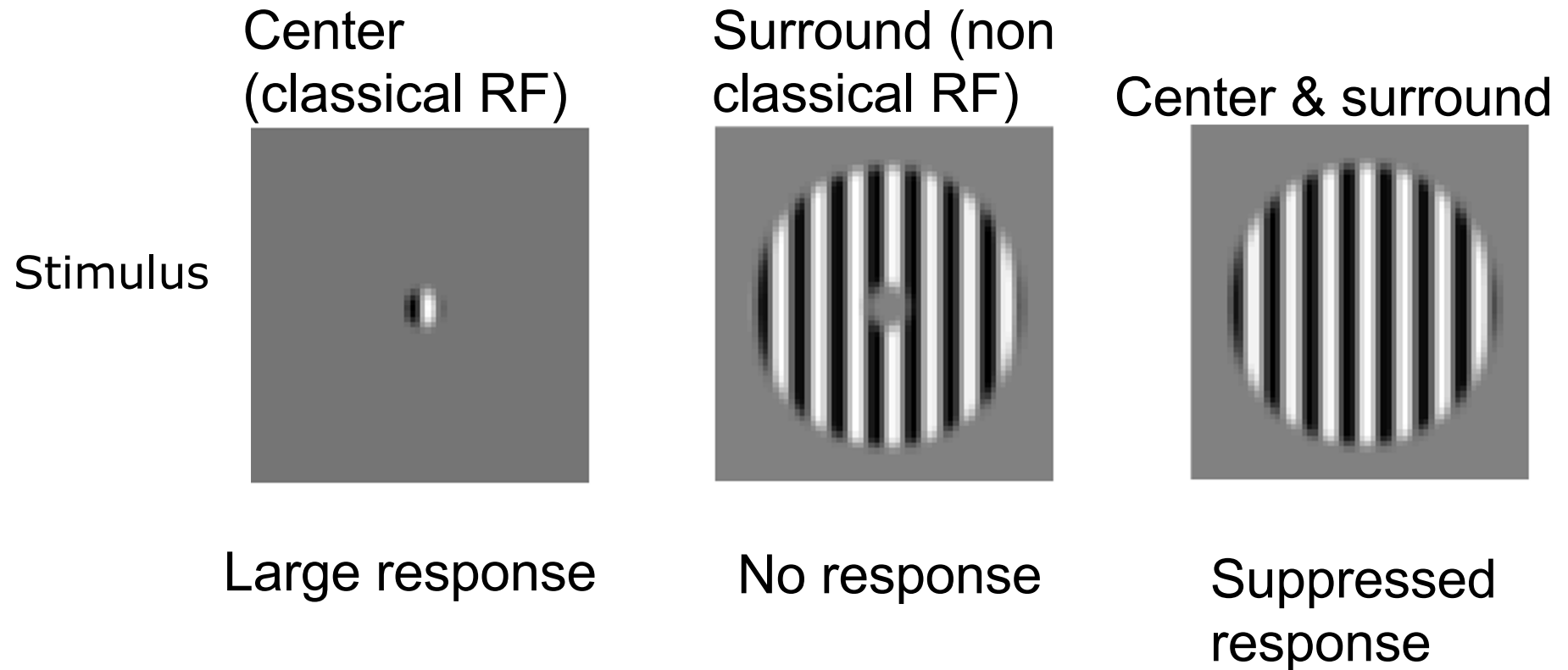
Surround (non  
classical RF)



No response

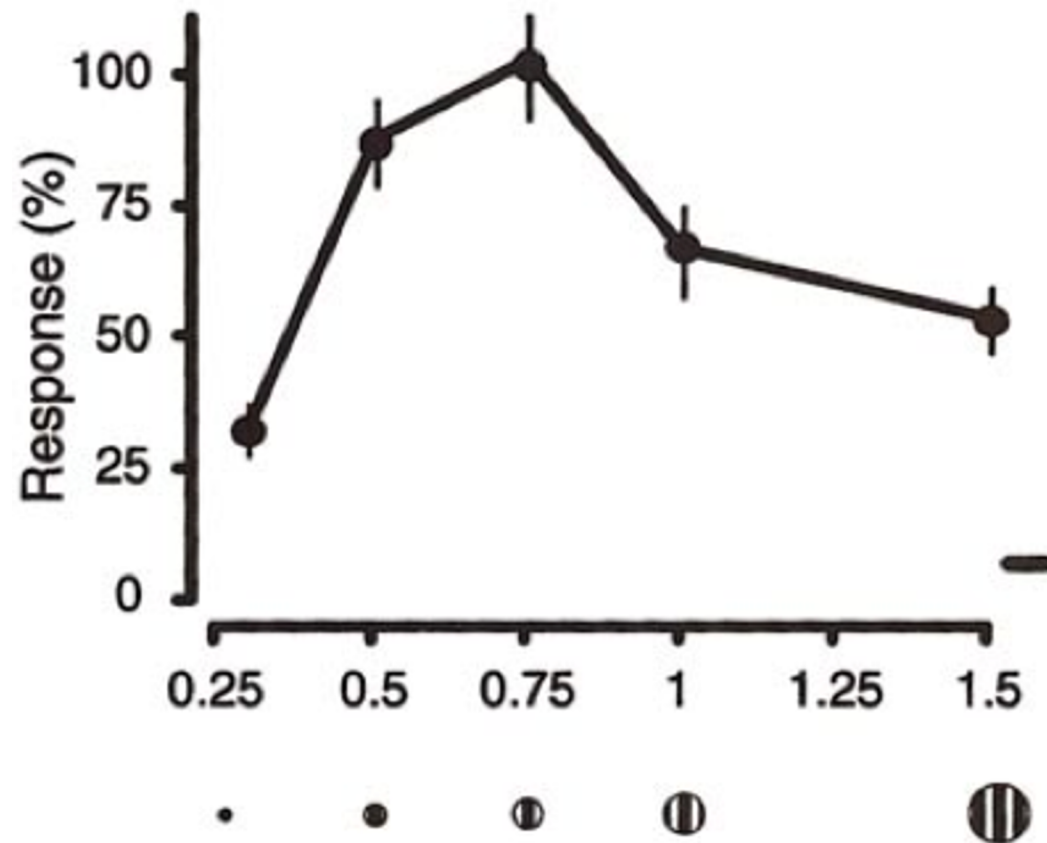
Surround stimulus defined such that by itself there is no response in the neuron

# Visual cortex: spatial surround



But surround stimulus can modulate response to center. Cortical neurons are affected by spatial context (often reduced response, as illustrated by spiking cartoon).

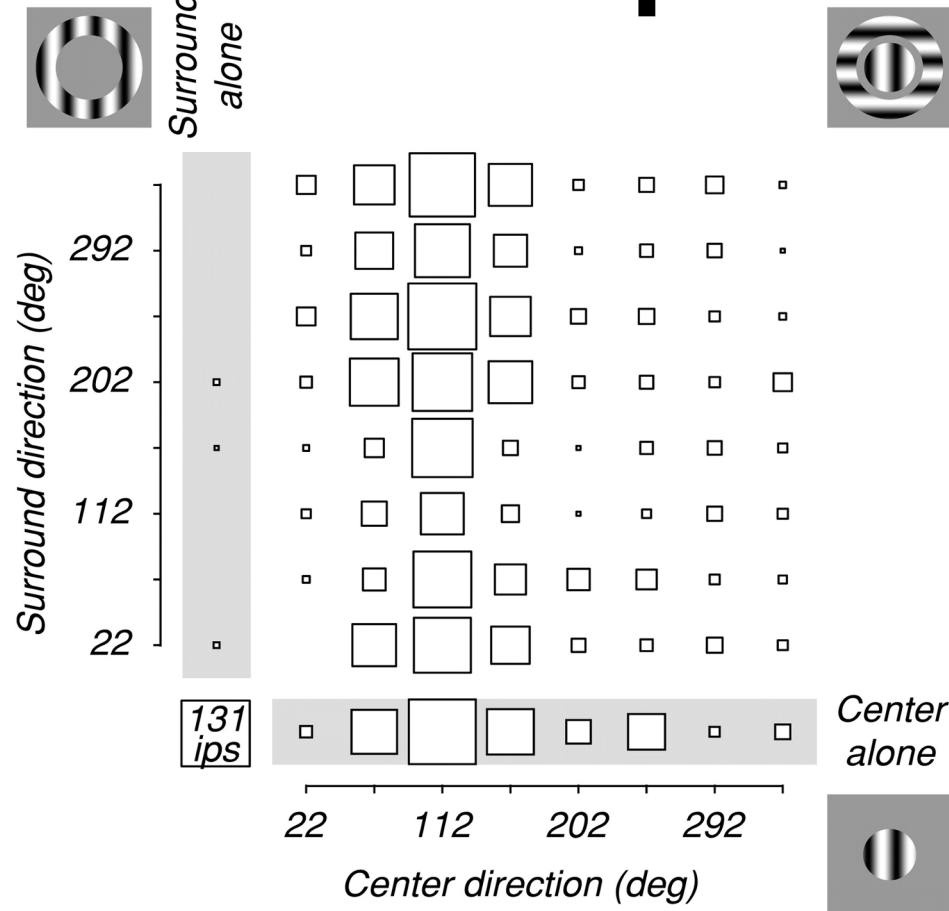
# Visual cortex: spatial surround



Jones and Sillito, 2001

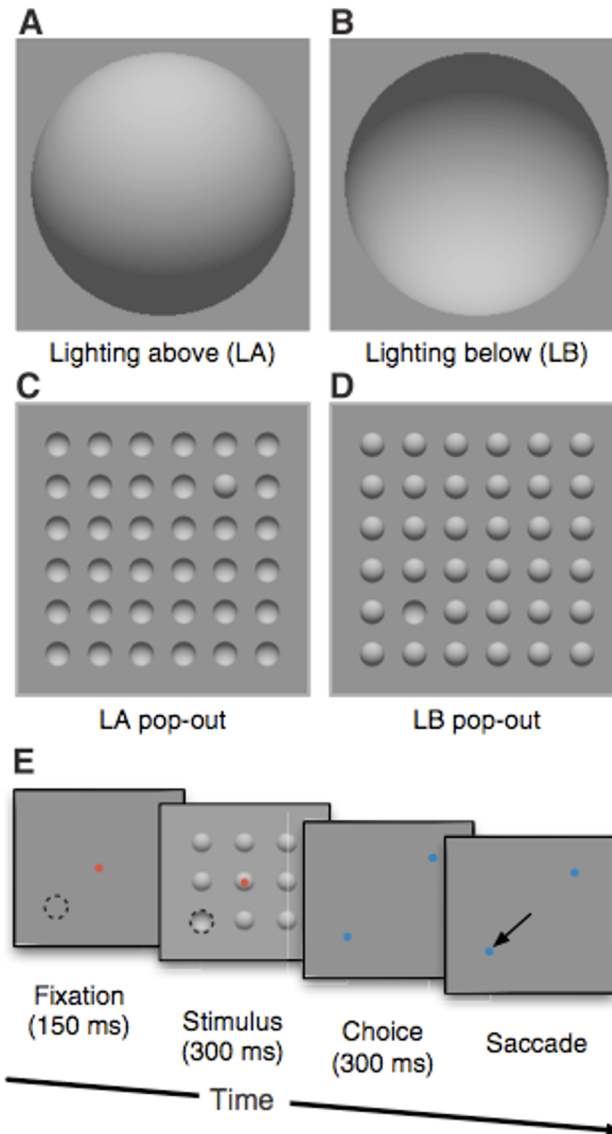
Response reduced as stimulus size (spatial context) made larger

# Visual cortex: spatial surround



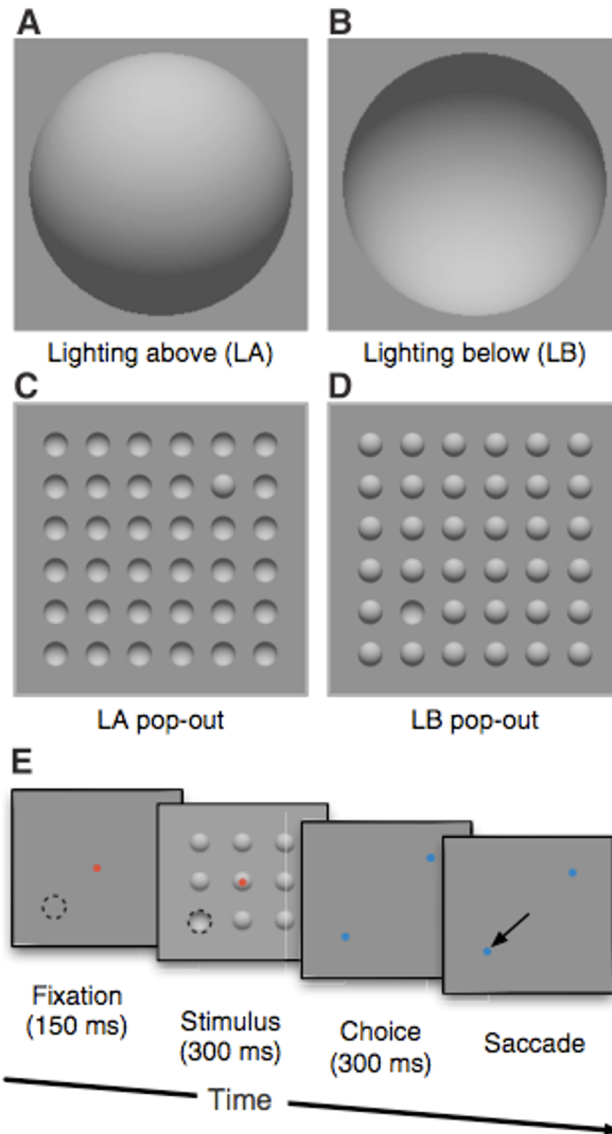
Response reduced most when the center and surround stimuli have the same orientation

# Context by other visual cues?



Smith et al. 2007

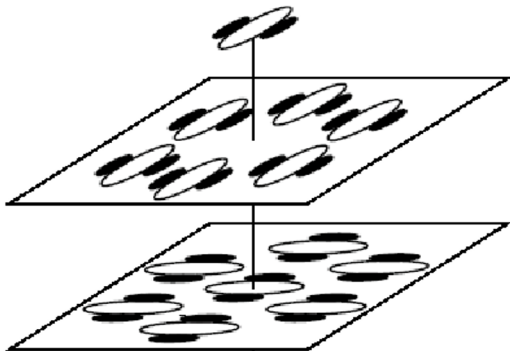
# Context by other visual cues?



Which one  
pops out more?

Smith et al. 2007

# Simple descriptive model of cortical surround effects



$$R_i = \frac{L_i^2}{\sum_j w_{ji} L_j^2 + \sigma^2}$$

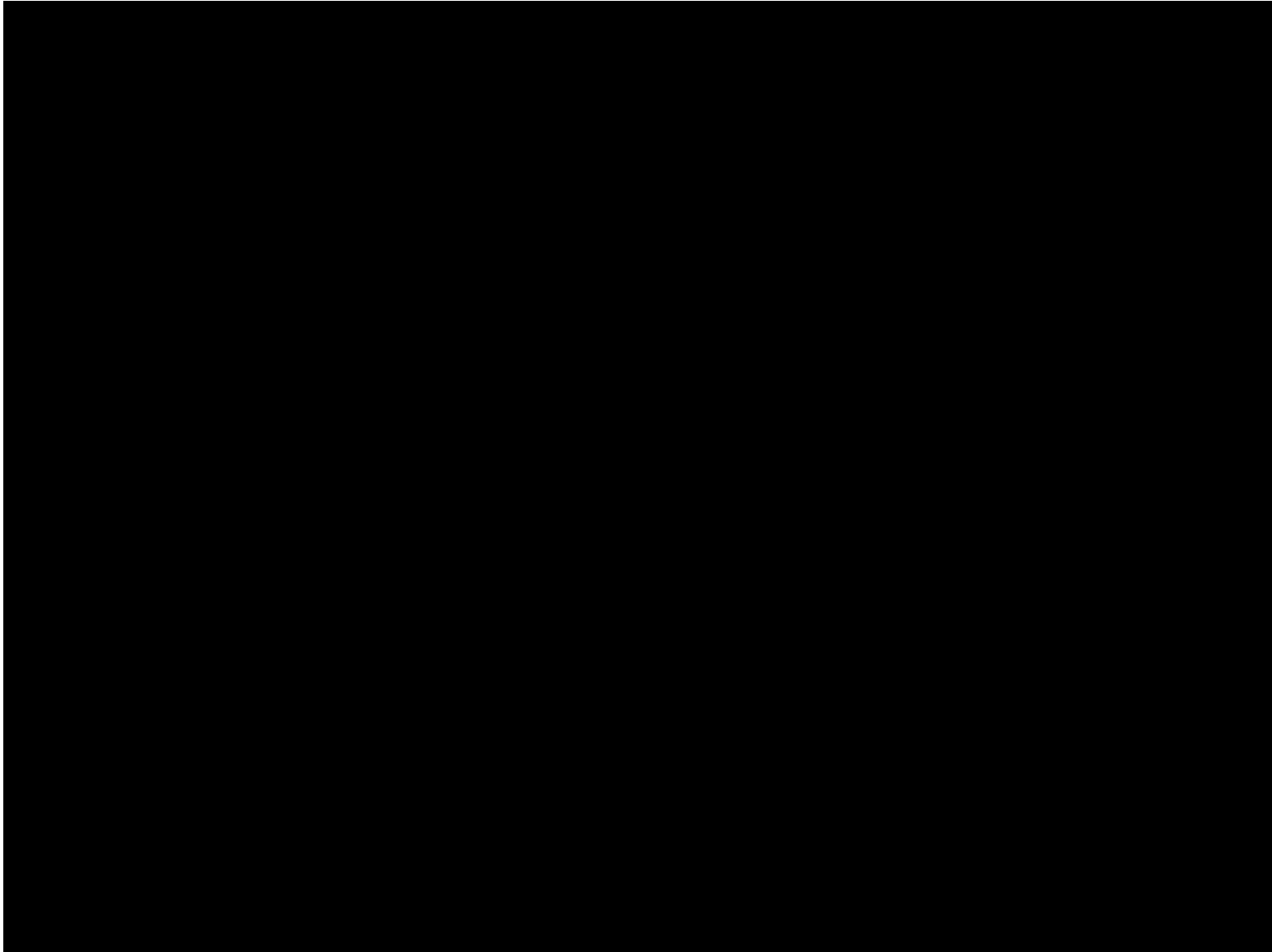
Linear filters followed by  
nonlinearity (divide by surround  
responses)

After Heeger 1992

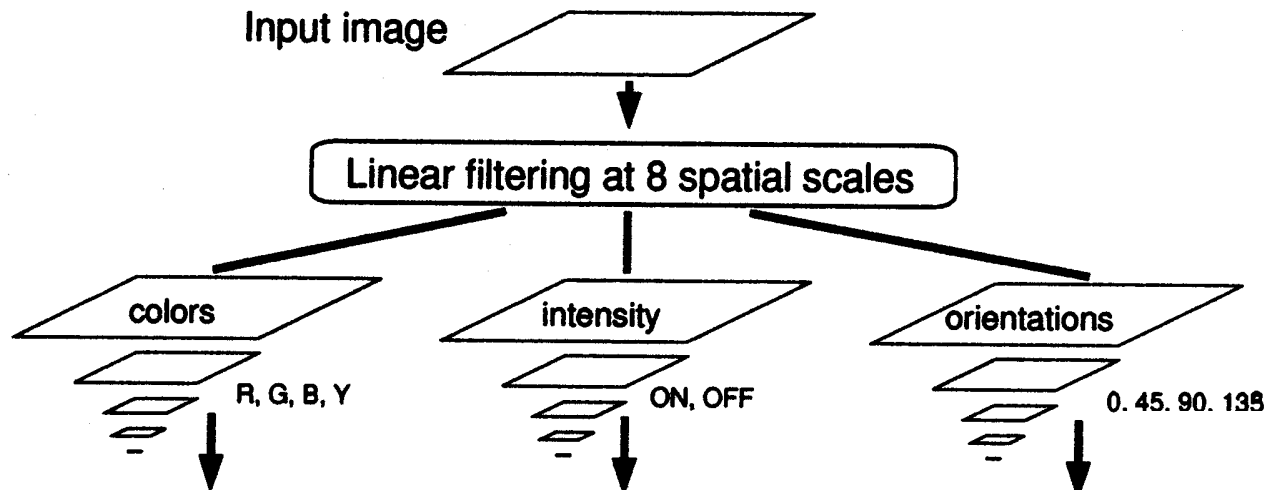


## **Eye movements and salience**

## **Example 1: Eye movements and salience (Laurent Itti, University of Southern California)**

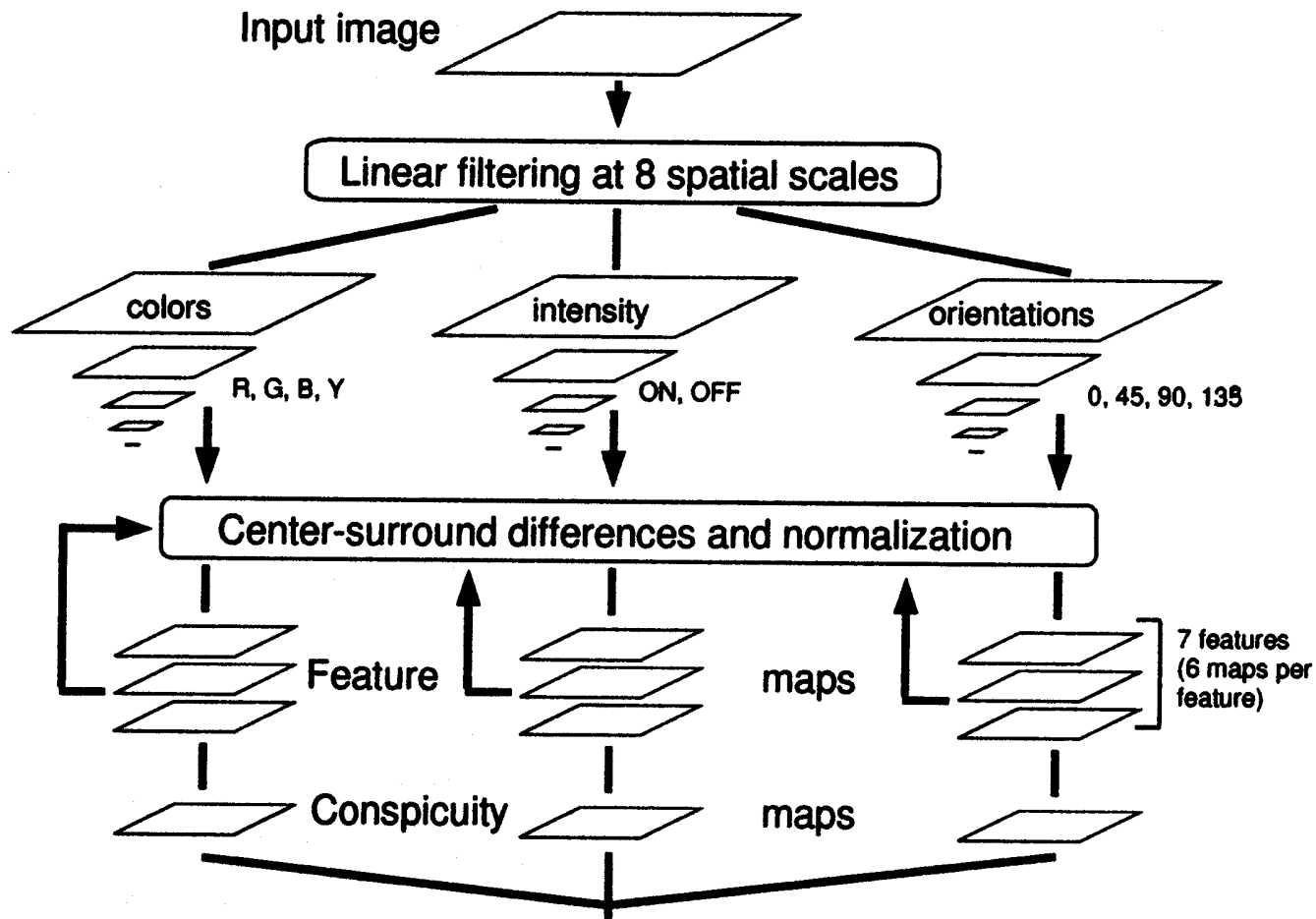


# Eye movements and saliency (Itti and Koch, 2000)



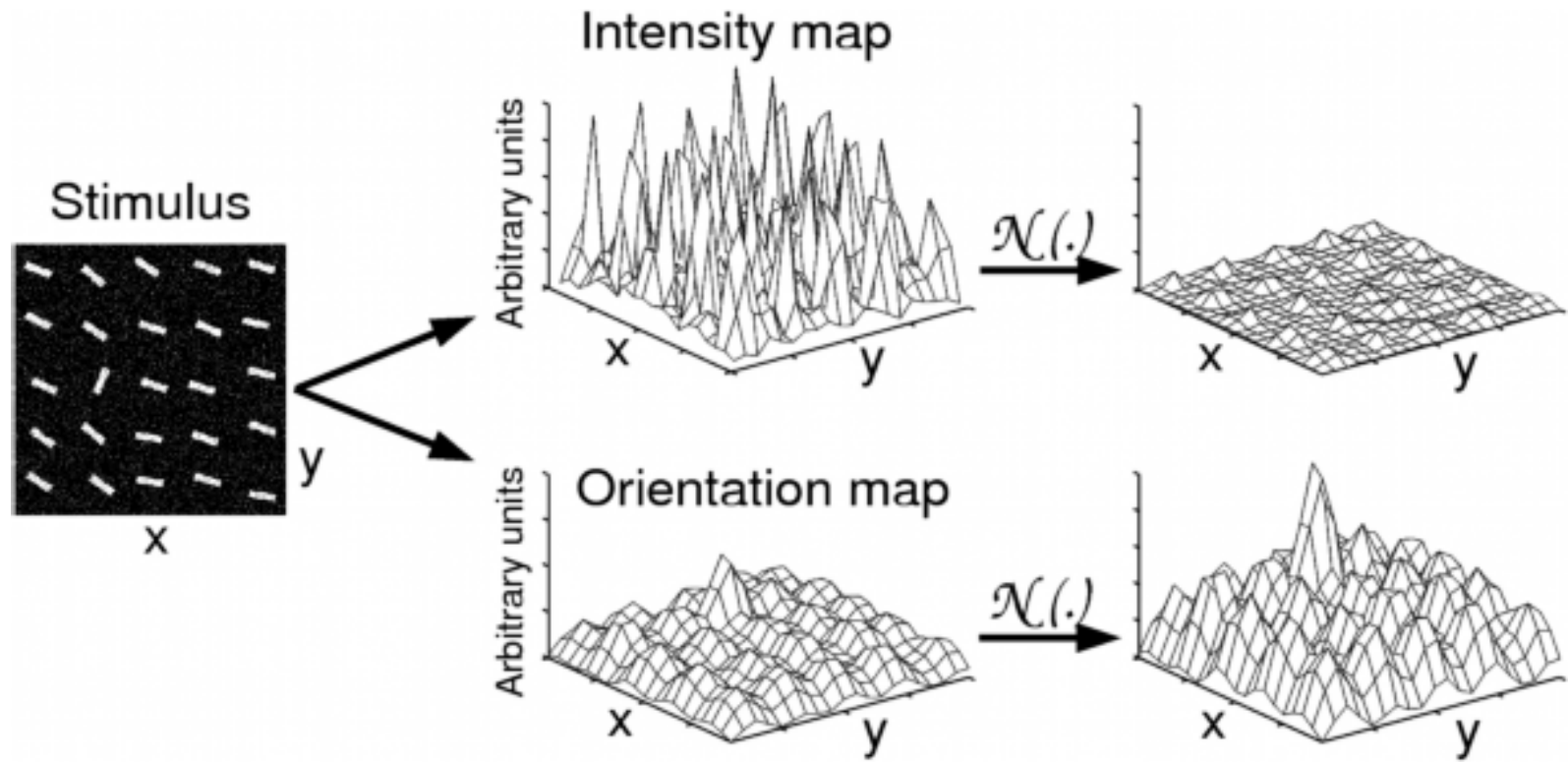
Analyze saliency for different features:  
Colors, intensity, orientations ....

# Eye movements and saliency (Itti and Koch, 2000)



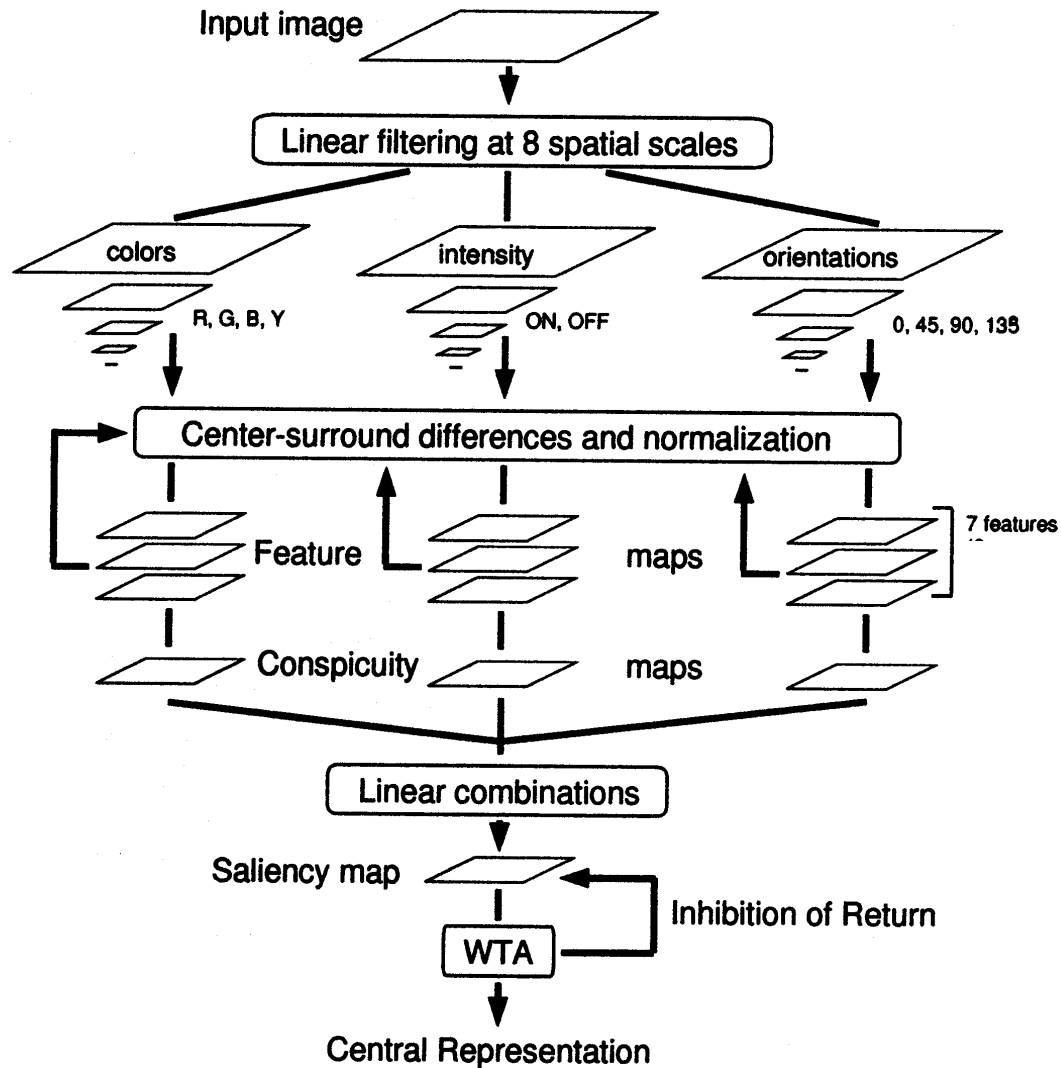
Analyze saliency for different feature maps:  
Colors, intensity, orientations ....

# Eye movements and salience (Itti and Koch, 2000)



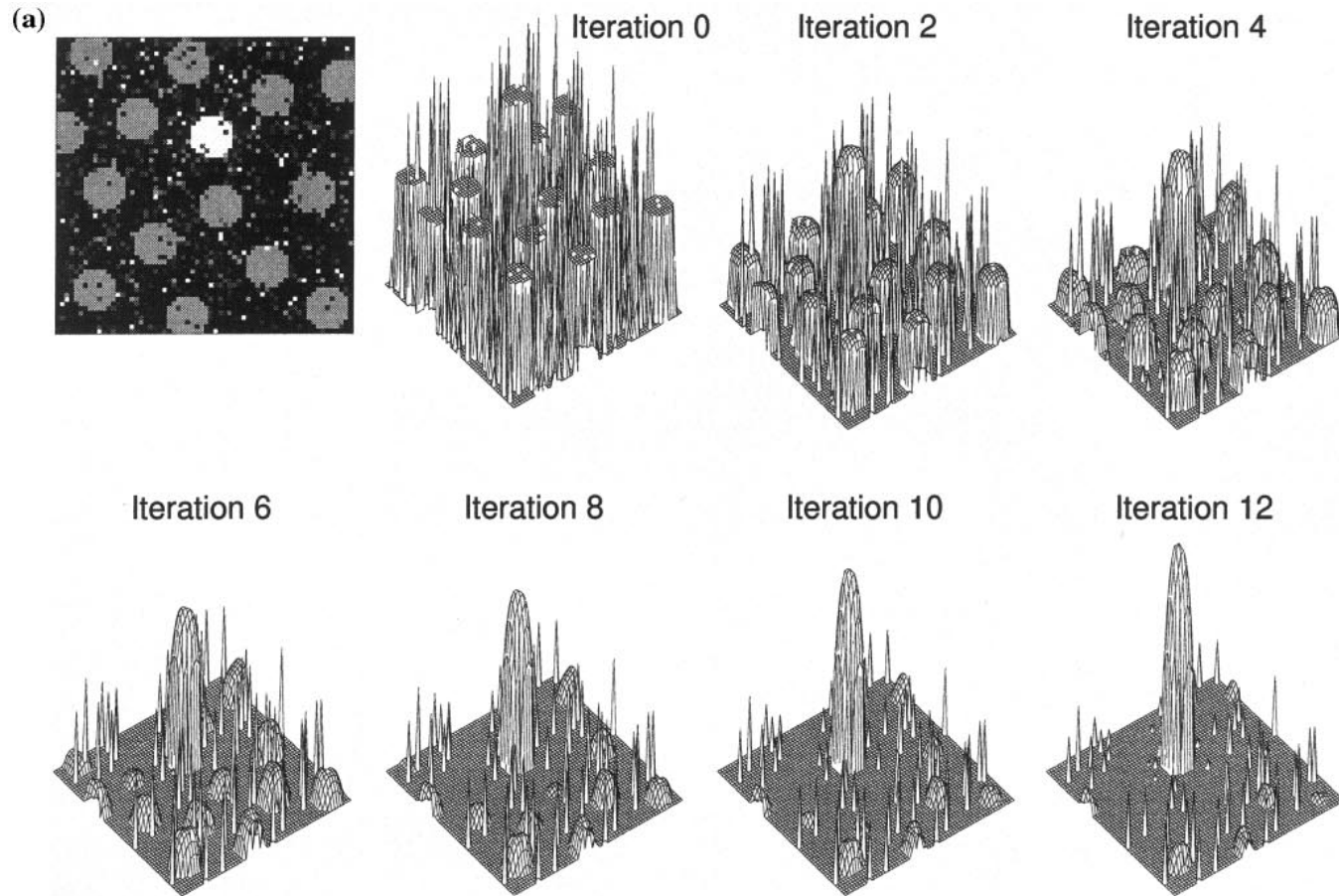
From Wikipedia page

# Eye movements and saliency (Itti and Koch, 2000)



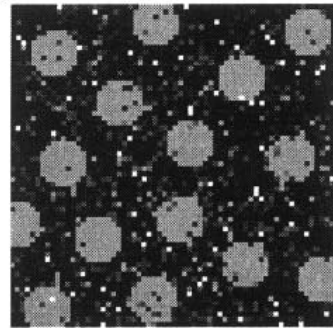
Linearly combine the saliency maps of different features

# Eye movements and saliency (Itti and Koch, 2000)

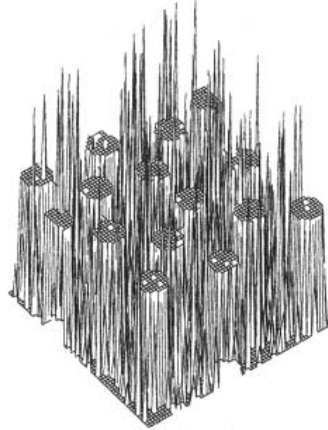


# Eye movements and saliency (Itti and Koch, 2000)

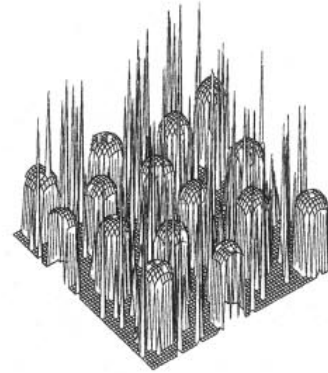
(b)



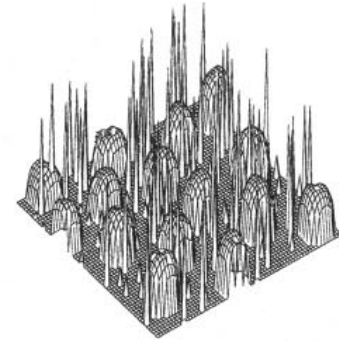
Iteration 0



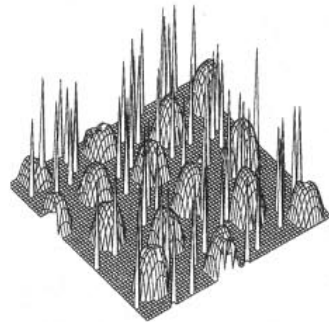
Iteration 2



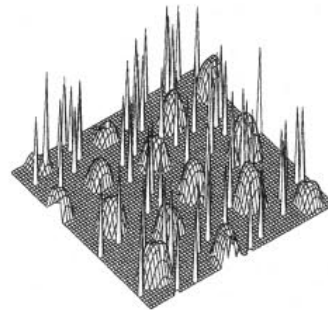
Iteration 4



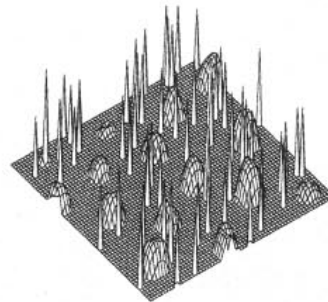
Iteration 6



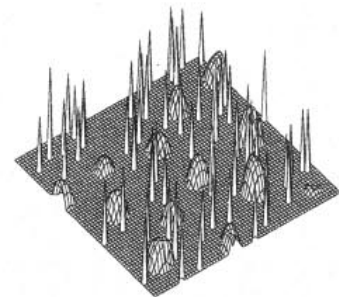
Iteration 8



Iteration 10

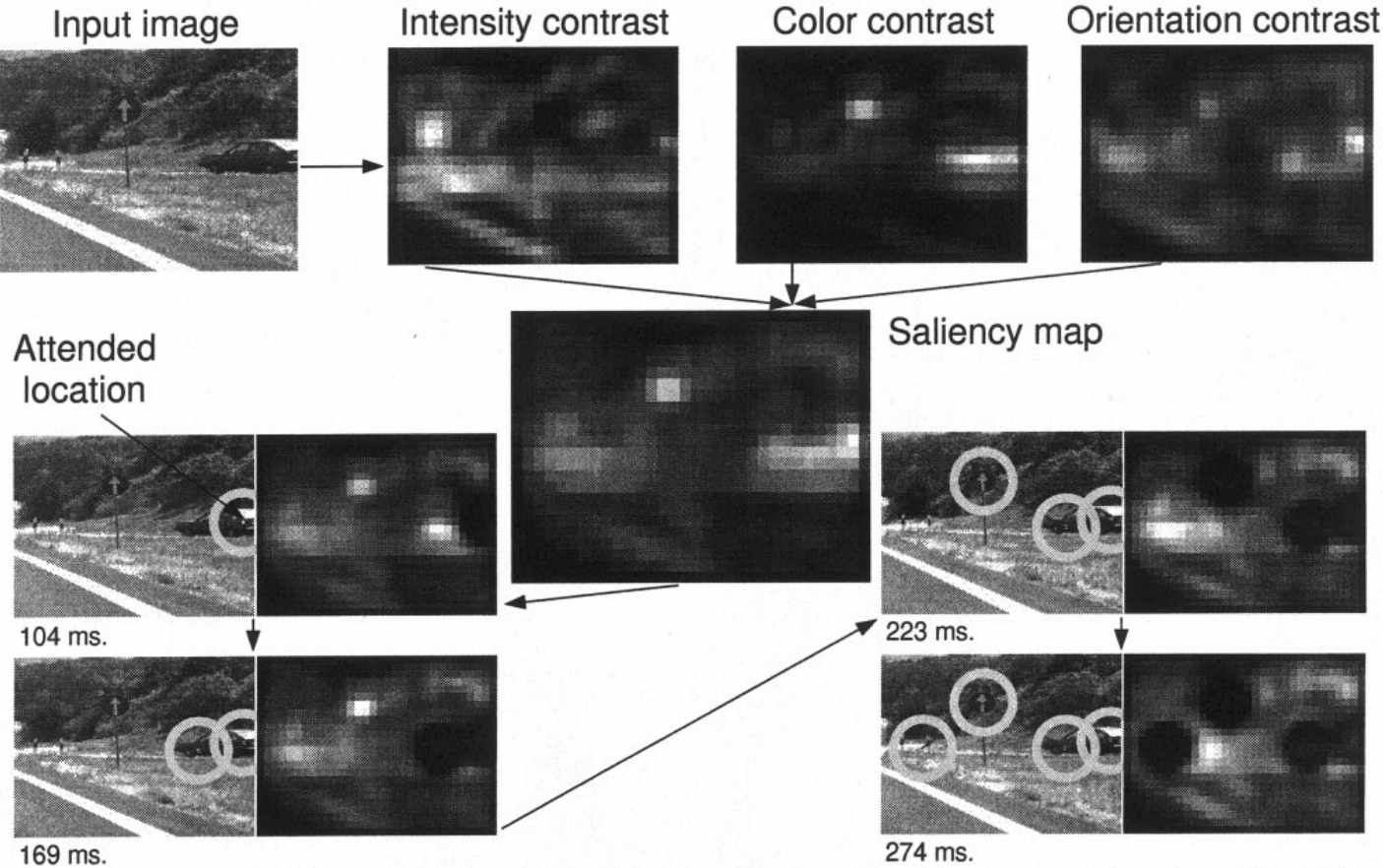


Iteration 12





# Eye movements and saliency (Itti and Koch, 2000)



# Eye movements: not only salience (Yarbus 1967)



## Eye movements: not only salience (Yarbus 1967)



Free examination



# Eye movements: not only salience (Yarbus 1967)



# Eye movements: not only salience (Yarbus 1967)



Remember the clothes worn by people



# Eye movements: not only salience (Yarbus 1967)



# Eye movements: not only salience (Yarbus 1967)



Give the ages of the people

# Eye movements: not only salience



Free examination.

1



Estimate material circumstances of the family

2



Give the ages of the people.

3



Surmise what the family had been doing before the arrival of the unexpected visitor.

4



Remember the clothes worn by the people.

5



Remember positions of people and objects in the room.

6



Estimate how long the visitor had been away from the family.

7

3 min. recordings of the same subject



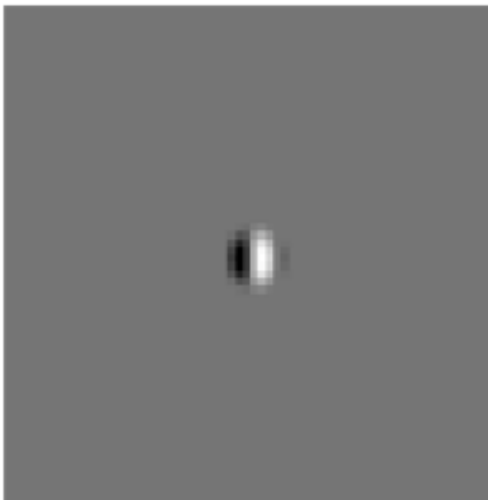
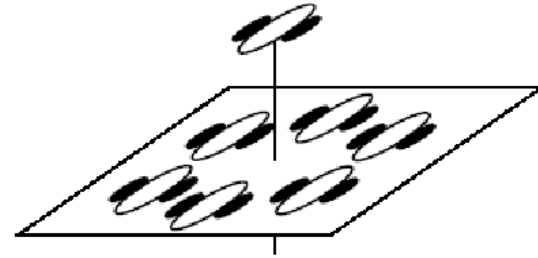
# **Surround spatial context and cortical neural processing of visual scenes**

# Motivation

- Spatial context plays critical role in object *grouping* and recognition, and in *segmentation*. It is key to everyday behavior; deficits have been implicated in neurological and developmental disorders and aging
- Poor understanding for how we (and our cortical neurons) process complex, natural images

# Contextual influences

- Cortical visual neurons (V1)



Large response



Suppressed response



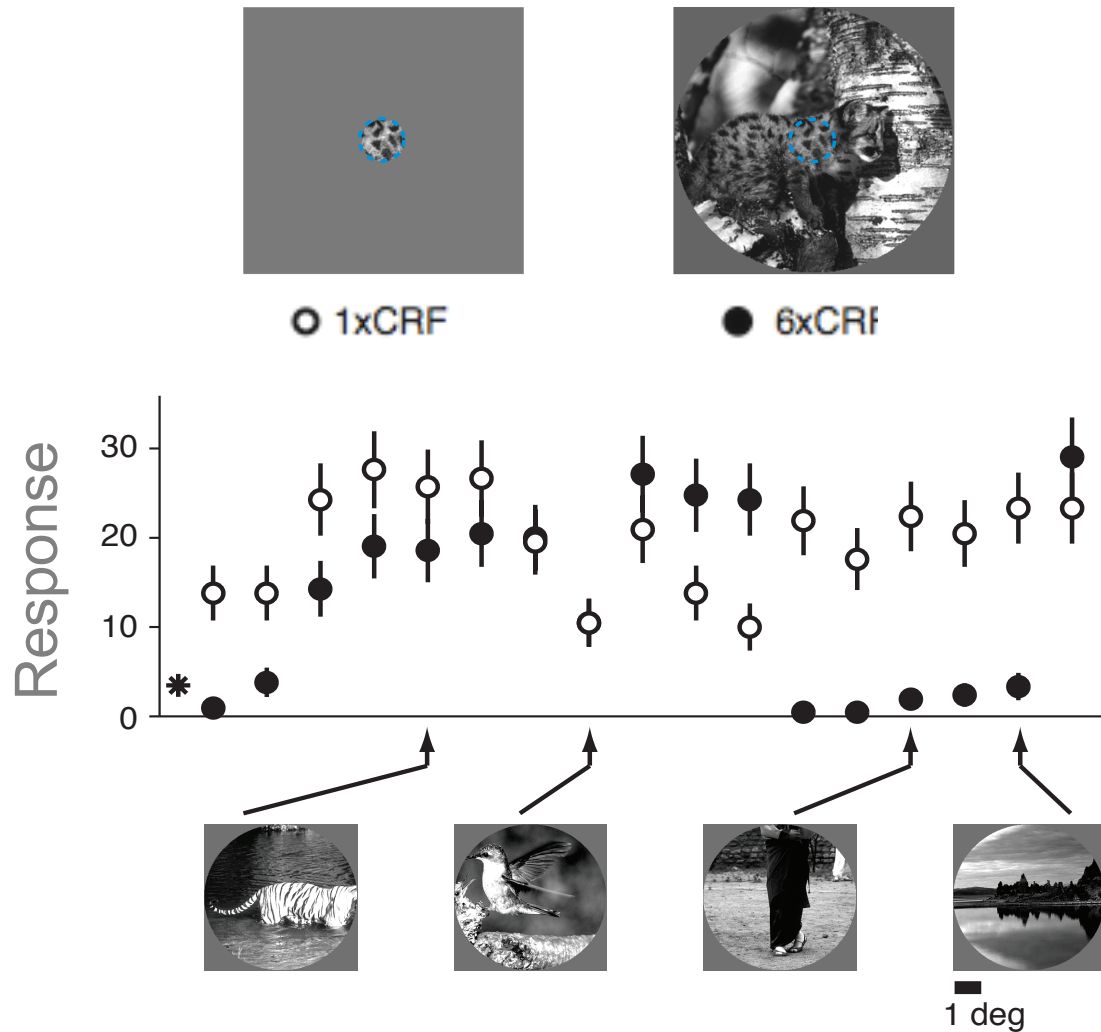
??

Simple oriented stimuli

Image

# Cortical Neurons

- Spatial context and natural scenes



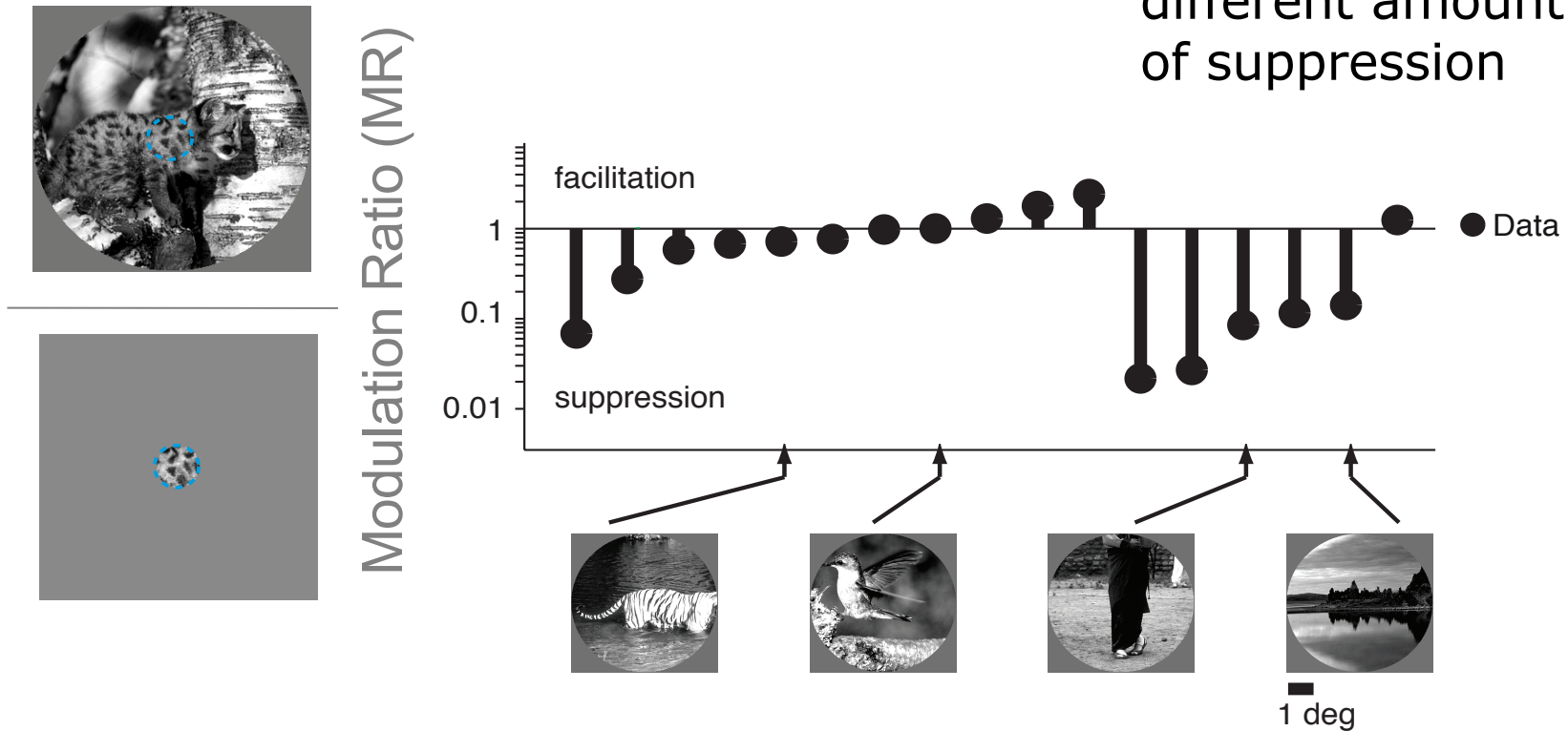
One neuron,  
different images,  
different amount  
of suppression  
by large stimuli

Data: Adam Kohn lab  
(Coen-Cagli, Kohn,  
Schwartz, 2015)

# Cortical Neurons

- Spatial context and natural scenes

One neuron,  
different images,  
different amount  
of suppression



Data: Adam Kohn lab (Coen-Cagli, Kohn, Schwartz, 2015)

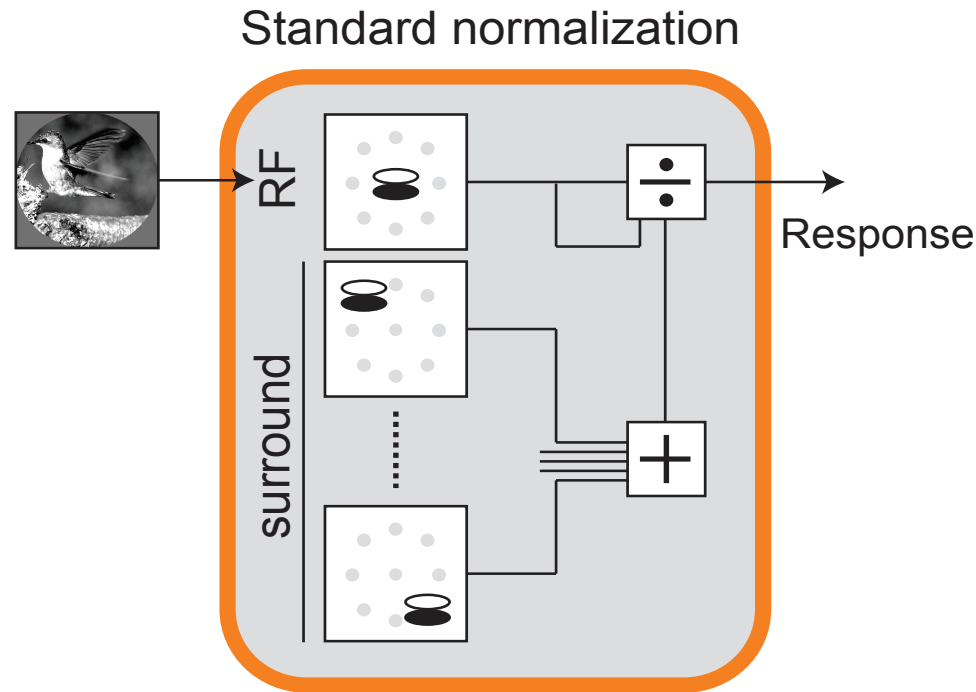
# Cortical Neurons

- Spatial context and natural scenes



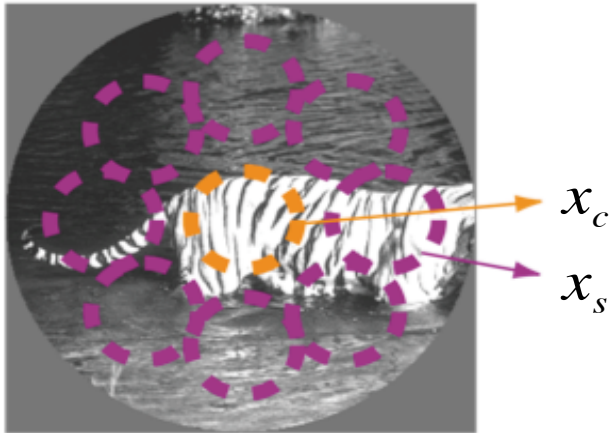
Can we capture data with  
**canonical** divisive normalization?  
(**descriptive model**)

# Divisive normalization



- Descriptive model
- Canonical computation (Carandini, Heeger, Nature Reviews Neuro, 2012)
- Has been applied to visual cortex, as well as other systems and modalities, multimodal processing, value encoding, etc
- Here center responses divided by surround responses

# Cortical Neurons



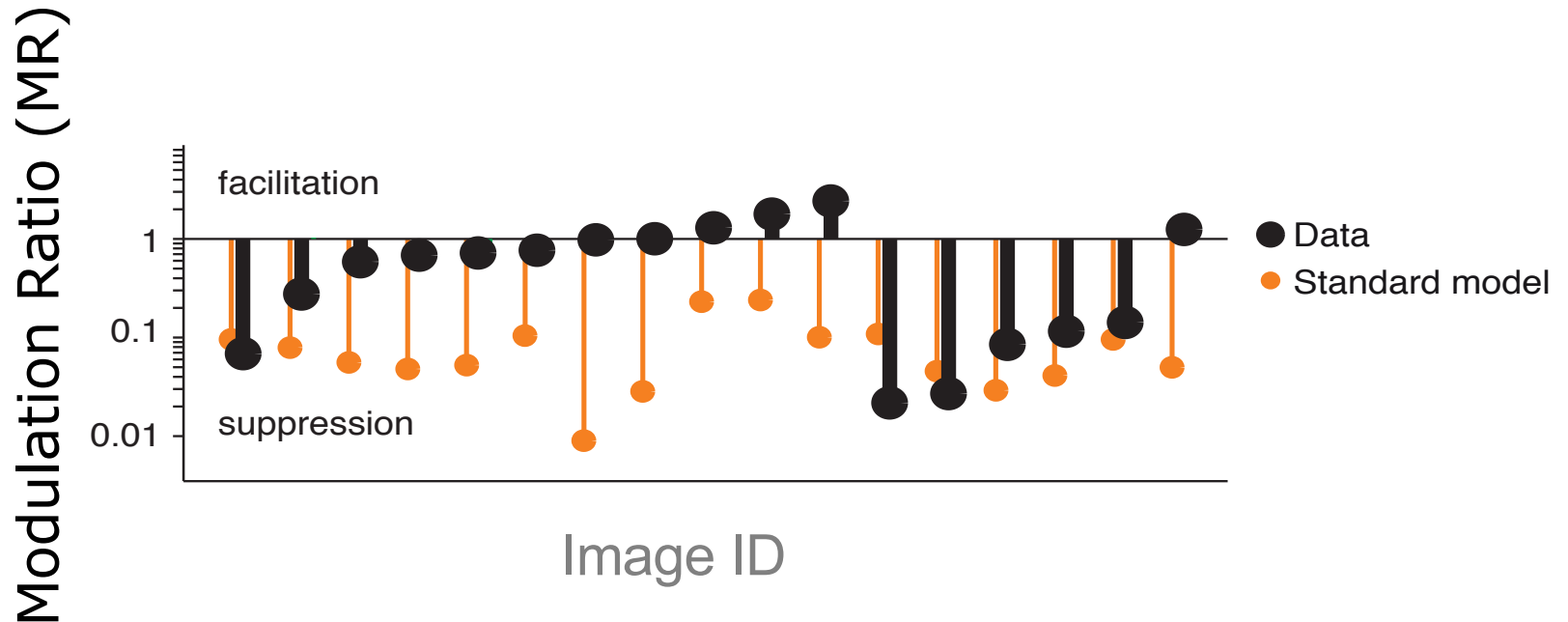
Canonical divisive normalization:

$$R_c \propto \frac{x_c}{\sqrt{x_c^2 + x_s^2}}$$

V1 Data: Adam Kohn lab



# Cortical responses to natural images

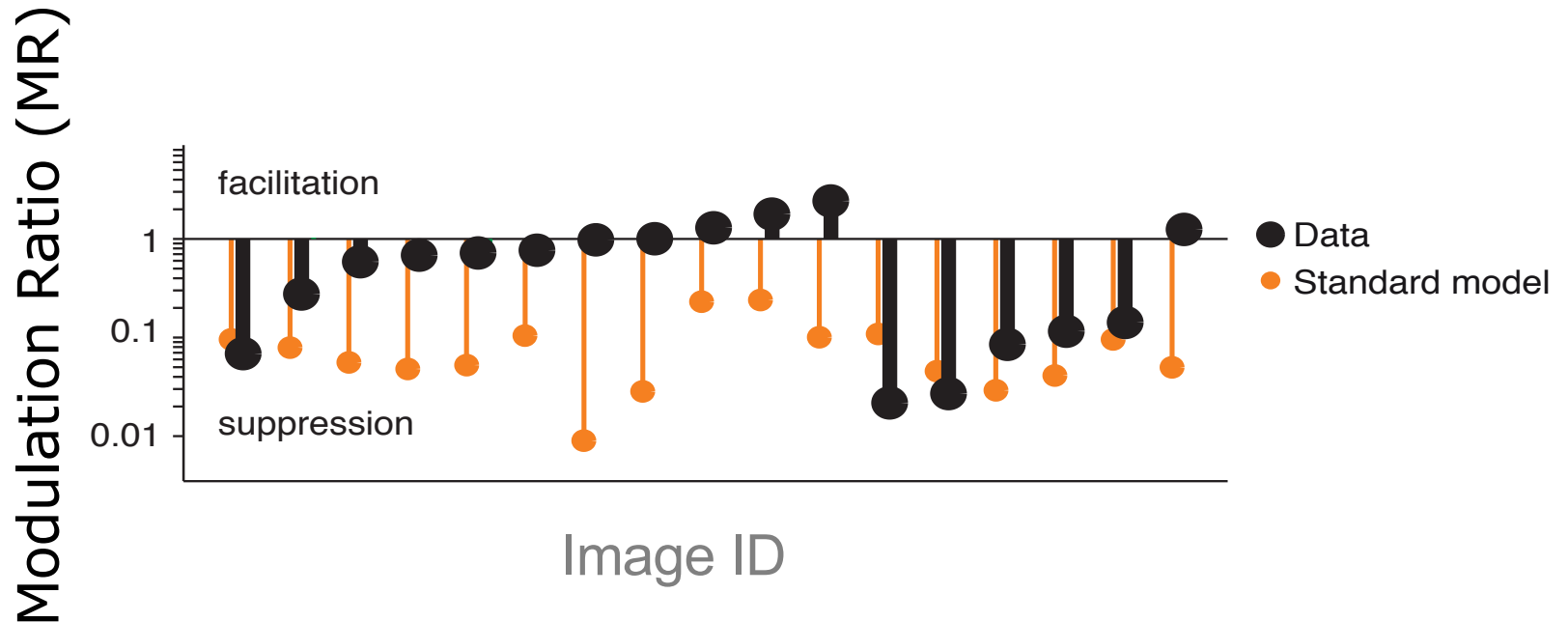


- We fit the standard normalization model to neural data
- Often predicts suppression when there is none in the data
- Poor prediction quality

Data: Adam Kohn lab

Coen-Cagli, Kohn, Schwartz, Nature Neuroscience 2015

# Cortical responses to natural images



- Can we explain as strategy to encode natural images optimally based on expected spatial contextual regularities?

Data: Adam Kohn lab

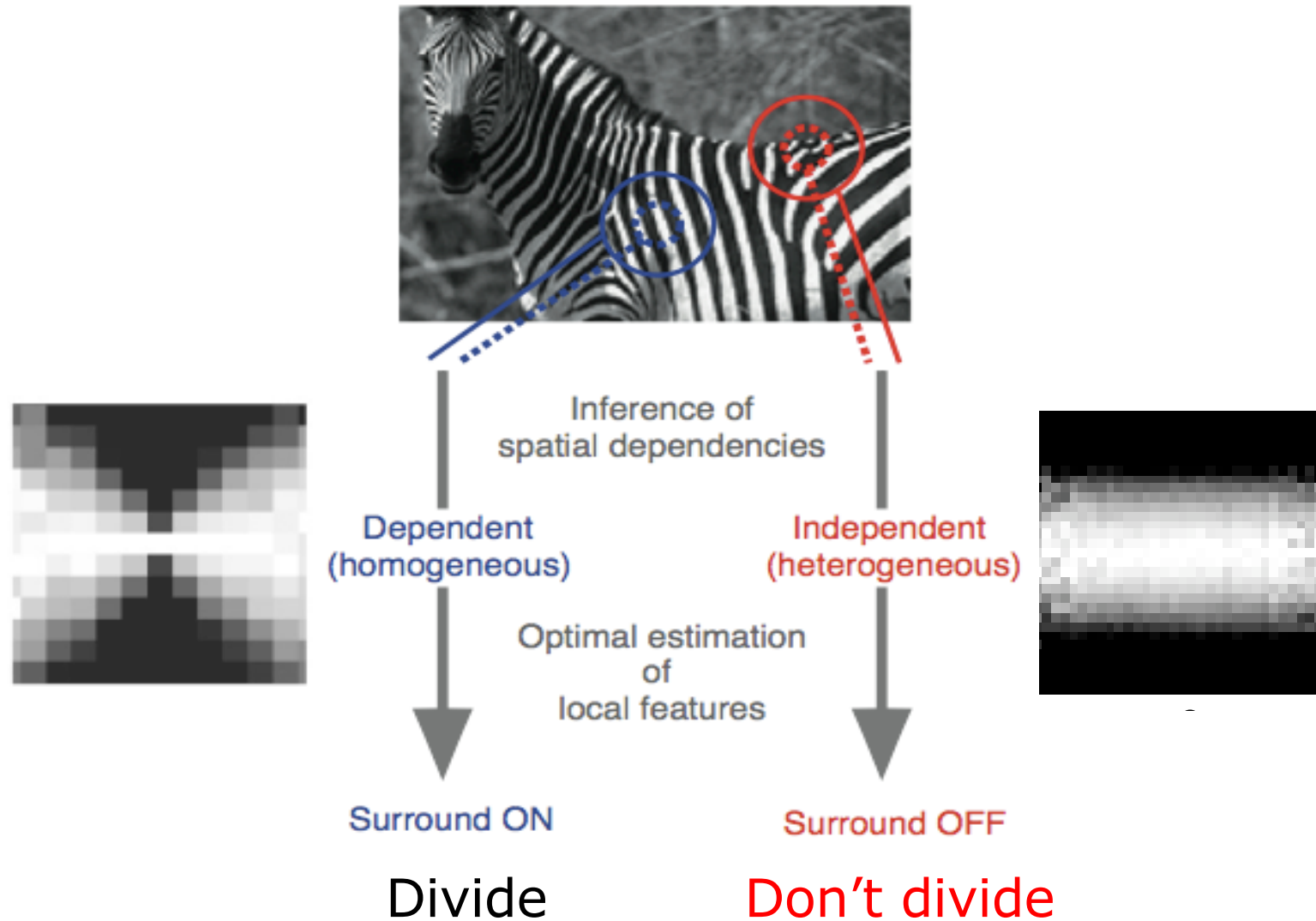
Coen-Cagli, Kohn, Schwartz, Nature Neuroscience 2015

## Divisive normalization: richer model



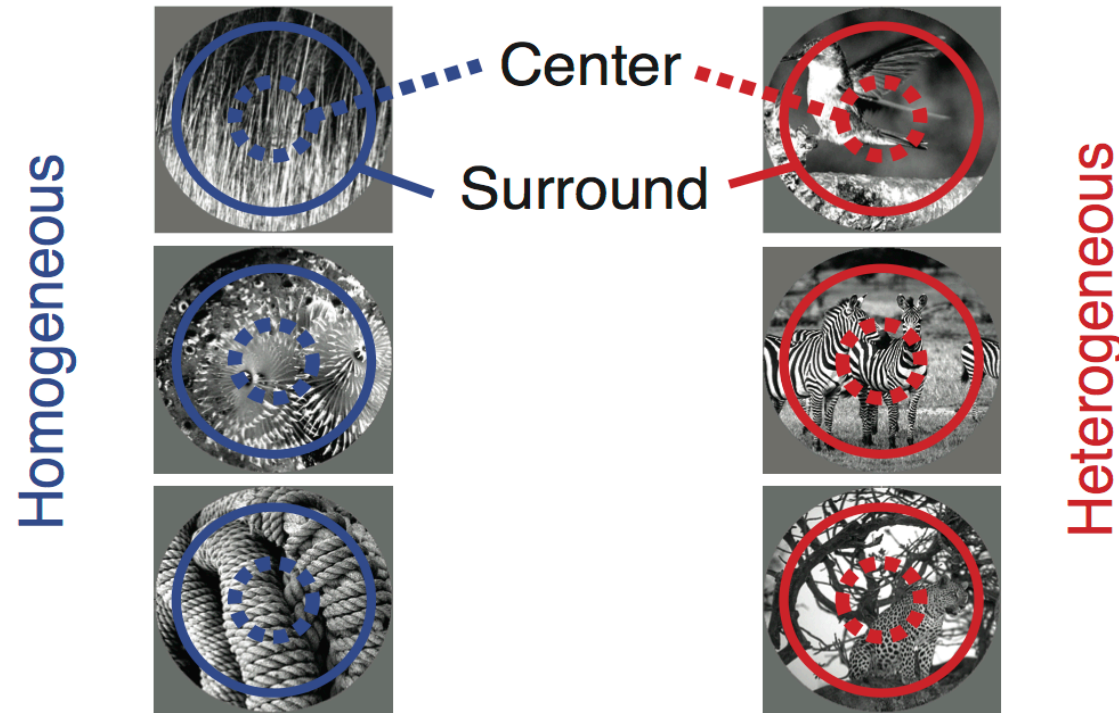
Divisive normalization *descriptive* models have been applied in many neural systems. We sought to develop a **richer model** based on image statistics

# Flexible Divisive Normalization



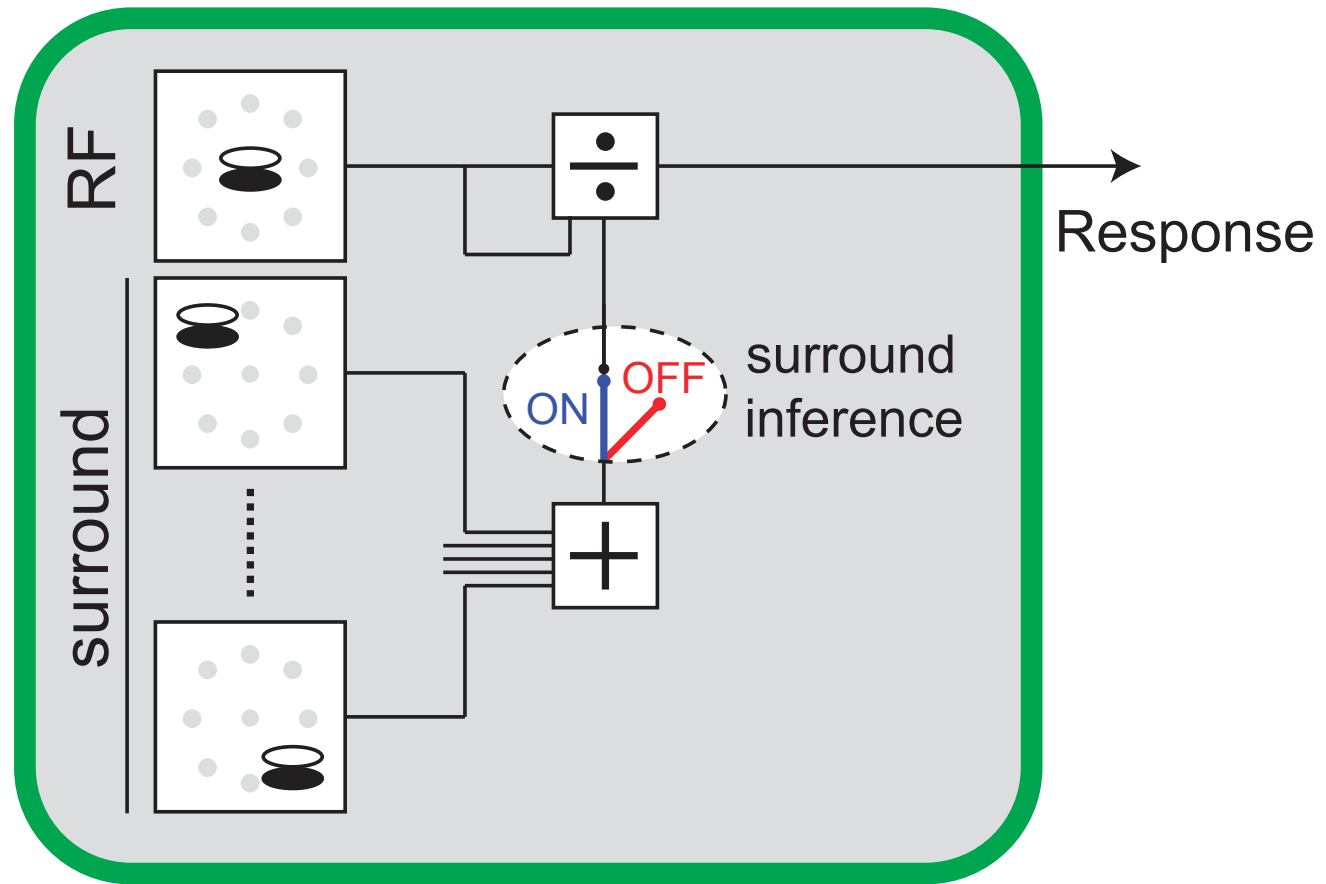
Model and experimental tests: Cagli, Kohn, Schwartz 2015

# Model predictions for natural images



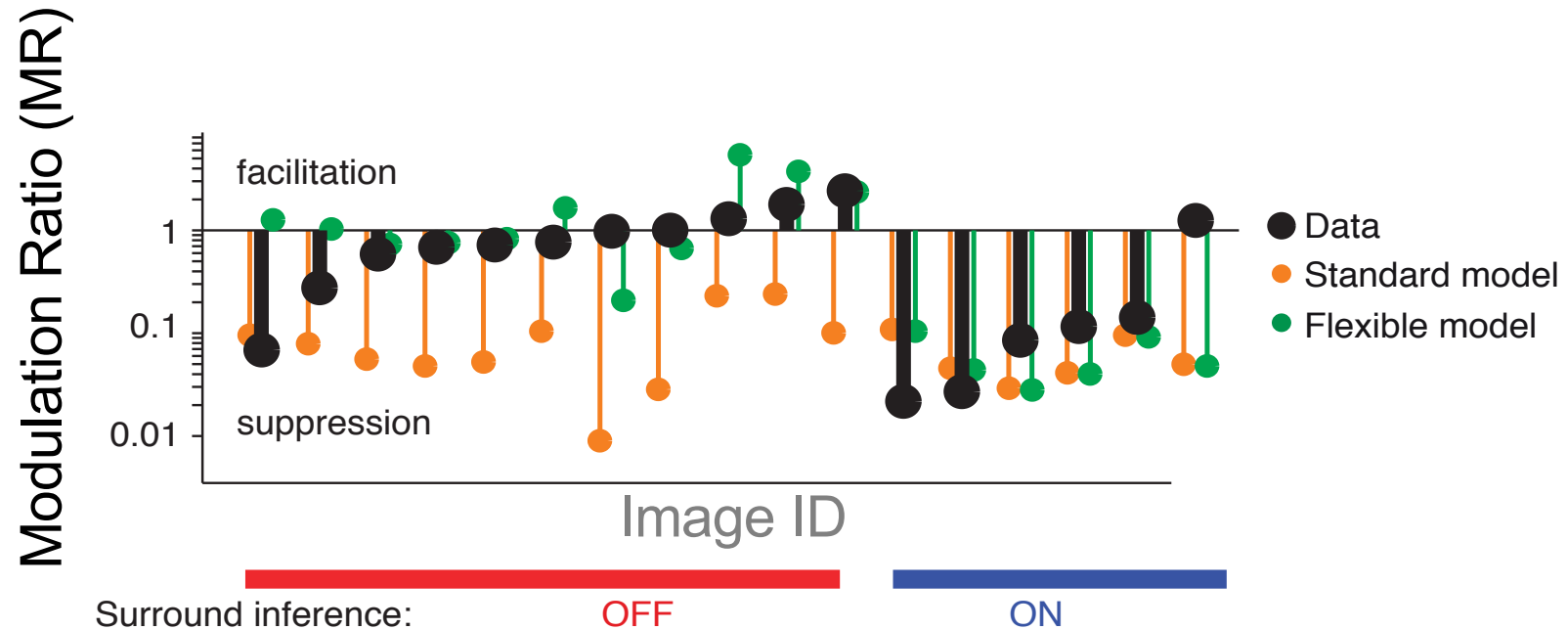
- **Homogeneous** and **heterogeneous** determined by model!
- Expect more suppression in neurons for homogeneous
- Related to salience (eg, Zhaoping Li)

# Model summary



Inference determined by model

# Natural scenes data



- Taking account of image statistics across space, we obtain better fit to neural data with the model

Coen-Cagli, Kohn, Schwartz, Nature Neuroscience, 2015

# Model Mechanisms

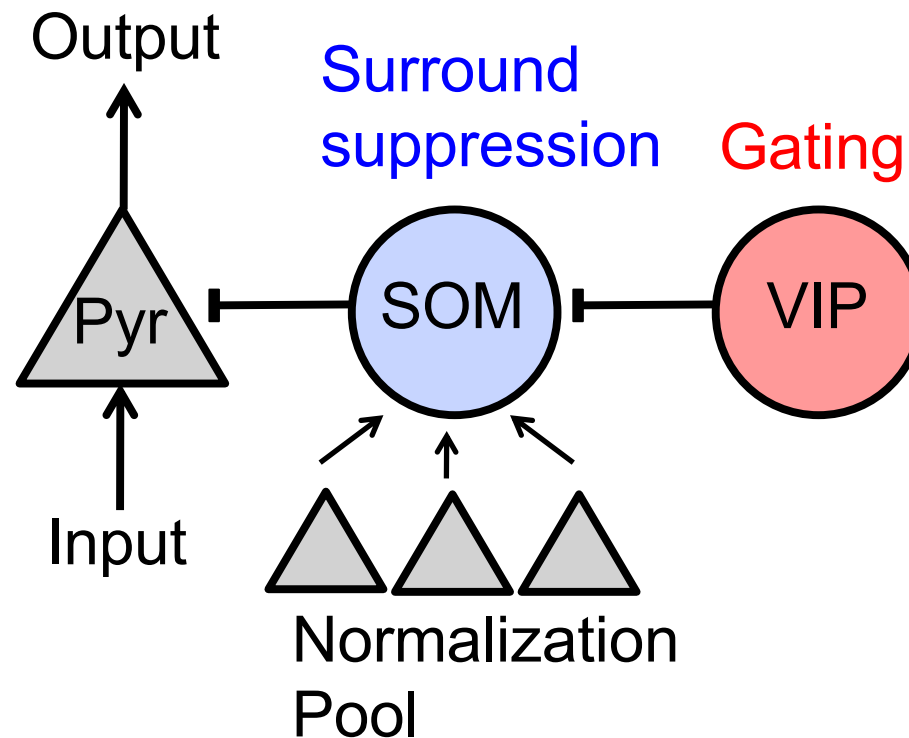
Divisive normalization:

- Feedback inhibition
- Distal dendrite inhibition
- Depressing synapses
- Internal biochemical adjustments
- Non-Poisson spike generation



# Flexible Normalization Mechanism?

- Adjusting gain by circuit mechanisms?
- Distinct classes of inhibitory interneurons? (eg, Adesnik, Scanziani et al. 2012; Pfeffer, Scanziani et al. 2013; Pi, Kepecs et al. 2013; Lee, Rudy et al. 2013)



# Key take-home points

- New approach to understanding cortical processing of natural images. Rather than fitting more complicated models, use insights from scene statistics
- Connects to neural computations that are ubiquitous, but enriches the “standard” model
- Our results suggest flexibility of contextual influences in natural vision, depending on whether center and surround are deemed statistically homogeneous

# Deep learning: normalization

Normalization has been shown to sometimes improve object recognition in deep neural networks

- Local normalization in Alexnet, 2012
- Other recent normalizations include: batch normalization in Ioffe and Szegedy, 2015; layer normalization in Ba et al., 2016
- More restricted than some of the normalizations used in cortical modeling
- But face some similar questions: How to choose what neural activations to normalize by