



HOW WE SEE ART AND HOW ARTISTS MAKE IT

Stephen Grossberg

Center for Adaptive Systems
Graduate Program in Cognitive and Neural Systems
Departments of Mathematics & Statistics, Psychological & Brain Sciences,
and Biomedical Engineering
Boston University



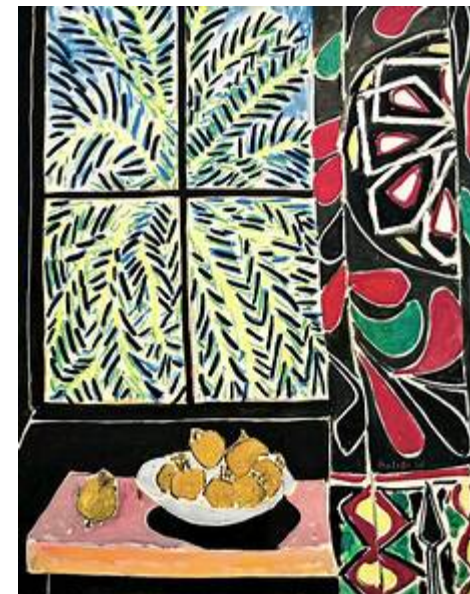
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S. Grossberg and L. Zajac, 2017,
Art & Perception, 5, 1-95
Published **OPEN ACCESS**
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**I will tell you a little about
HOW OUR BRAINS SEE
and discuss
how various painters struggled to
intuitively understand how THEY see
to
generate desired aesthetic effects
in their paintings**

HOW CAN A TALK ON THIS TOPIC EVEN BE GIVEN?

The results are based on the most advanced neural models of
HOW OUR BRAINS SEE

The models emerged through 40 years of research

They also offer an explanation of
what goes on in each brain
as it consciously sees, hears, feels, or knows something

See

Grossberg, S. (2017). Towards solving the hard problem of consciousness:
The varieties of brain resonances and the conscious experiences that they
support, *Neural Networks*, 87, 38-95

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and on my web page

WHAT IS THIS TALK ABOUT?

Paintings of visual artists activate multiple brain processes that contribute to conscious perception

Paintings of different artists and artistic movements emphasize different combinations of brain processes to achieve their aesthetic goals

Neural models of how advanced brains see characterize these processes, and were used to analyze paintings of 10 painters:

Jo Baer, Banksy, Ross Bleckner, Gene Davis, Charles Hawthorne, Henry Hensche, Henri Matisse, Claude Monet, Jules Olitski, Frank Stella

**How were such models discovered
in the first place?**

**How were such models discovered
in the first place?**

**How much do we know today about
how our brains work?**

HOW DOES A BRAIN GIVE RISE TO A MIND?

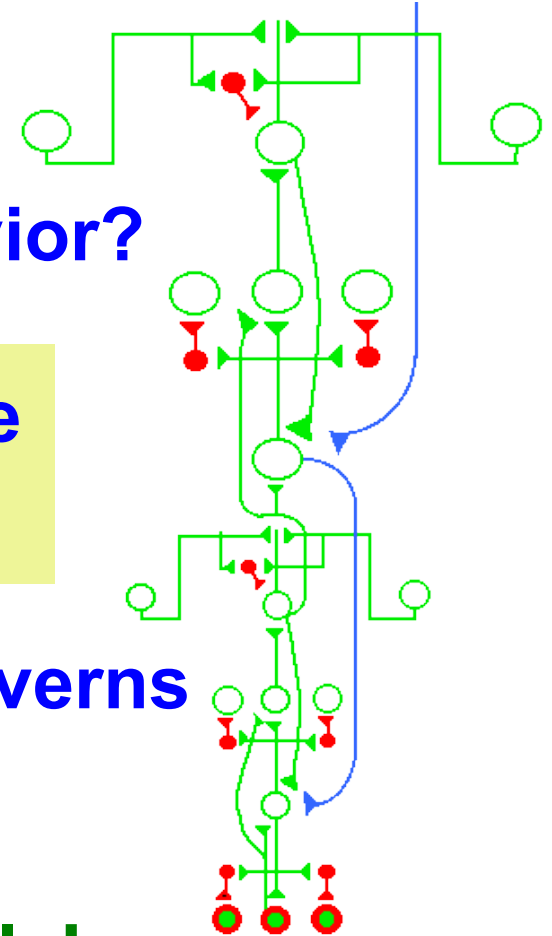
What **level** of brain organization controls behavior?

What is the **functional unit** of behavior?

BRAIN evolution needs to achieve **BEHAVIORAL** success

What level of **BRAIN** processing governs **BEHAVIORAL** success?

The **NETWORK** and **SYSTEM** levels!



How does **BEHAVIOR** arise as **EMERGENT PROPERTIES** of cells interacting in **NEURAL NETWORKS**?

Does this mean that individual neurons are unimportant?
Not at all!

How are individual **NEURONS** designed and connected in **NETWORKS** whose emergent properties give rise to successful **BEHAVIORS**?

Need to simultaneously describe **3 levels** (at least):

BEHAVIOR
NETWORK
NEURON

and a **MODELING** language to link them

...one reason why our brains are so hard to understand

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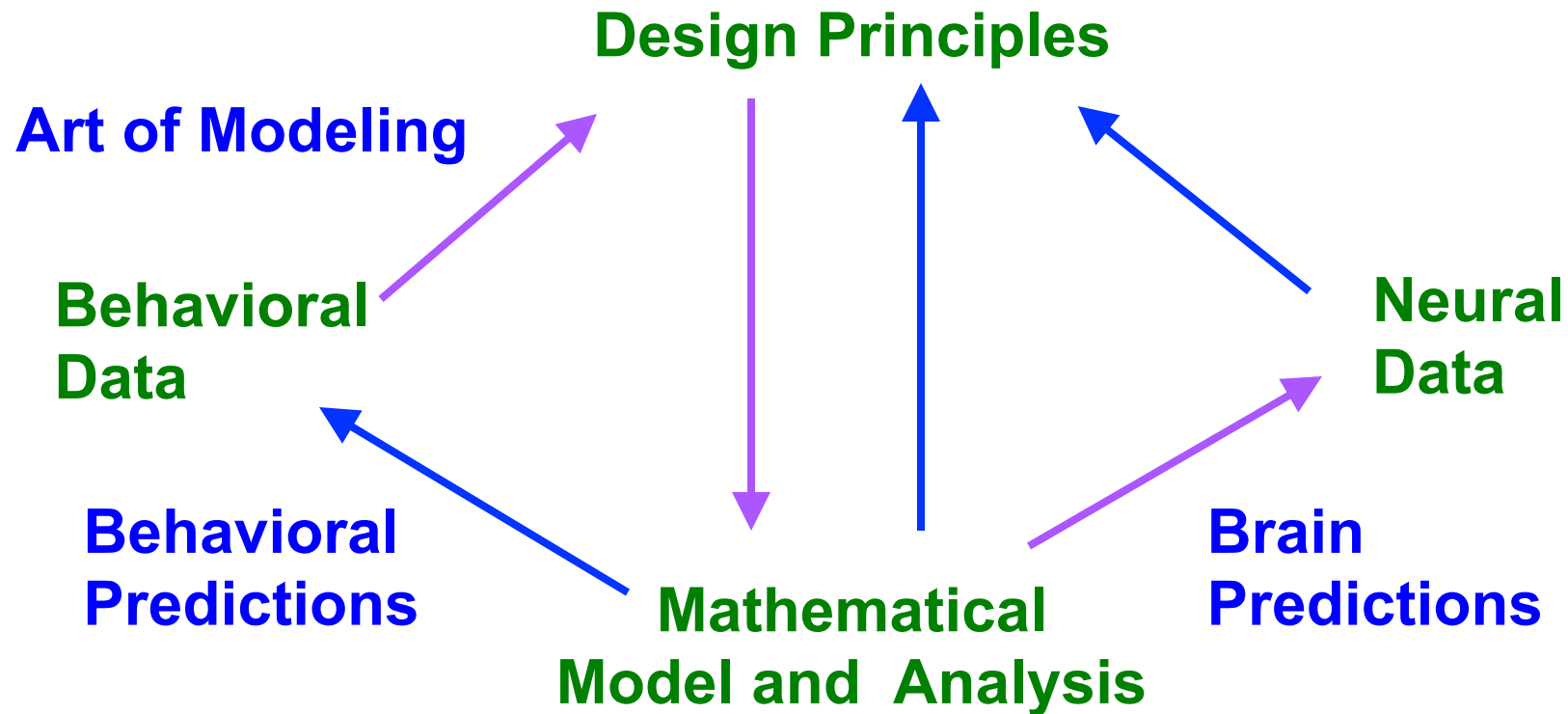
MAIN CONCLUSION

To deeply understand how brains work, you need to understand **HOW *EVOLUTION* SELECTS BRAIN DESIGNS** based on their **BEHAVIORAL SUCCESS**

To do this, start with **BEHAVIORAL DATA** that describe the behavior-environment interactions that are selected by *evolution*

Starting bottom-up with **ONLY** facts about neurons has not achieved an understanding of brain function for any complex behavior

MODELING METHOD AND CYCLE



Embedding Principle: Repeat this cycle to carry out model evolution, progressively unlumping the model to achieve increasing model realism and explanatory power

**AFTER GOING THROUGH THE MODELING CYCLE,
WHAT'S THE RESULT?!**

**IS THE BRAIN JUST A
“BAG OF TRICKS”?**



Ramachandran, 1990

NO!

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TRUE THEORIES ARE EMERGING

A small number of equations

e.g., shunting activation dynamics (STM)
habituated transmitter gates (MTM)
activity-gated learning (LTM) ...

A larger number of modules*

e.g., on-center off-surround nets
resonant matching nets
opponent processing nets
spectral timing nets
boundary completion nets
filling-in nets...

Specialized combinations of modules*, using a few basic equations, are assembled in architectures that solve modal problems

A still larger number of modal architectures

e.g. vision
audition
smell
touch
cognition
emotion...

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Modules are microassemblies, not the “independent modules” of AI

WHAT PRINCIPLES DETERMINE HOW MODAL ARCHITECTURES ARE DESIGNED?

BREAKTHROUGHS IN BRAIN COMPUTING

Models that link detailed **BRAIN CIRCUITS** to the
ADAPTIVE BEHAVIORS that they control

Mind/Body Problem

Describe **NEW PARADIGMS** for brain computing

INDEPENDENT MODULES
Computer Metaphor

COMPLEMENTARY COMPUTING

What is the nature of brain specialization?

LAMINAR COMPUTING

Why are all neocortical circuits organized in layers?

How do laminar circuits give rise to biological intelligence?

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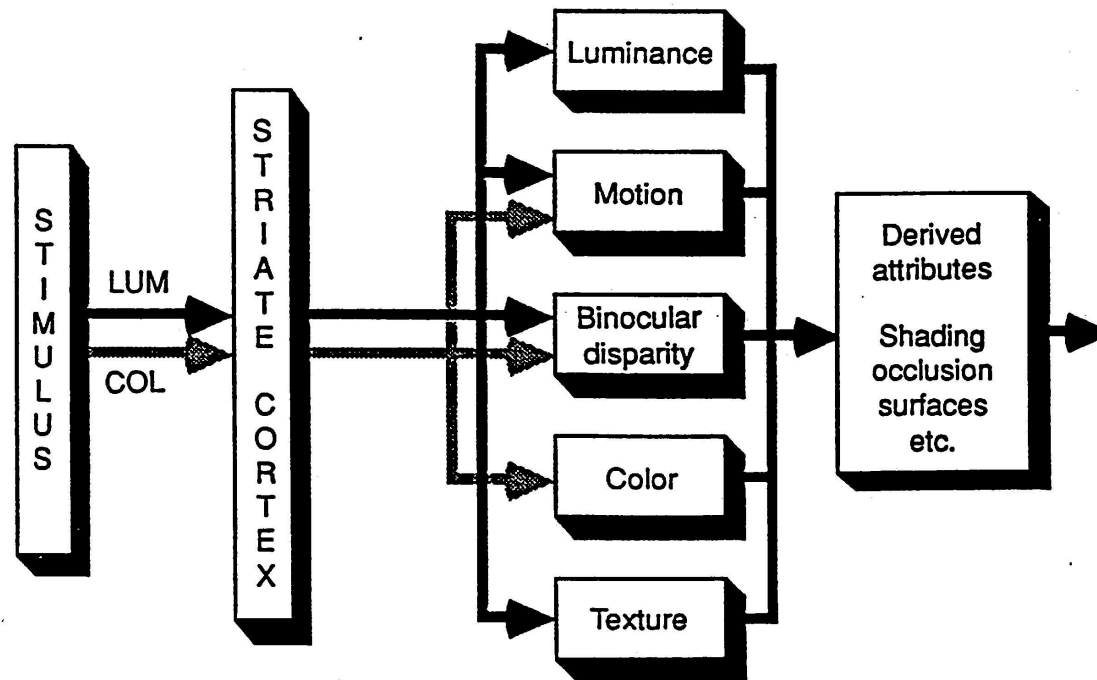
What is the nature of brain specialization?

LAMINAR COMPUTING

Why are all neocortical circuits organized in layers?

How do laminar circuits give rise to biological intelligence?

INDEPENDENT MODULES ARE APPEALING, BUT...?



The appeal: Lots of specialized brain regions in the visual cortex

Specialization does not, however, imply independent modules

INDEPENDENT modules should compute each property by itself

However, huge databases show interactions between (e.g.)
brightness and depth, motion and color, motion and depth,
texture and depth, texture and motion...**RELEVANT TO ART!**

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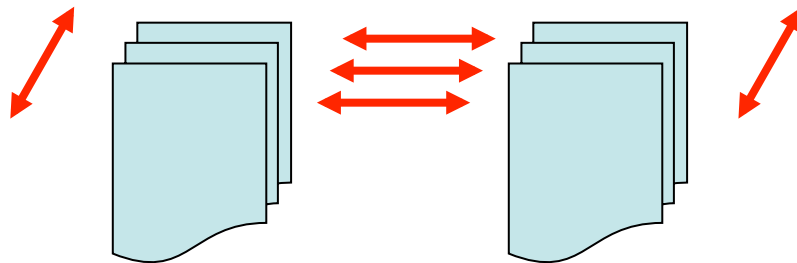
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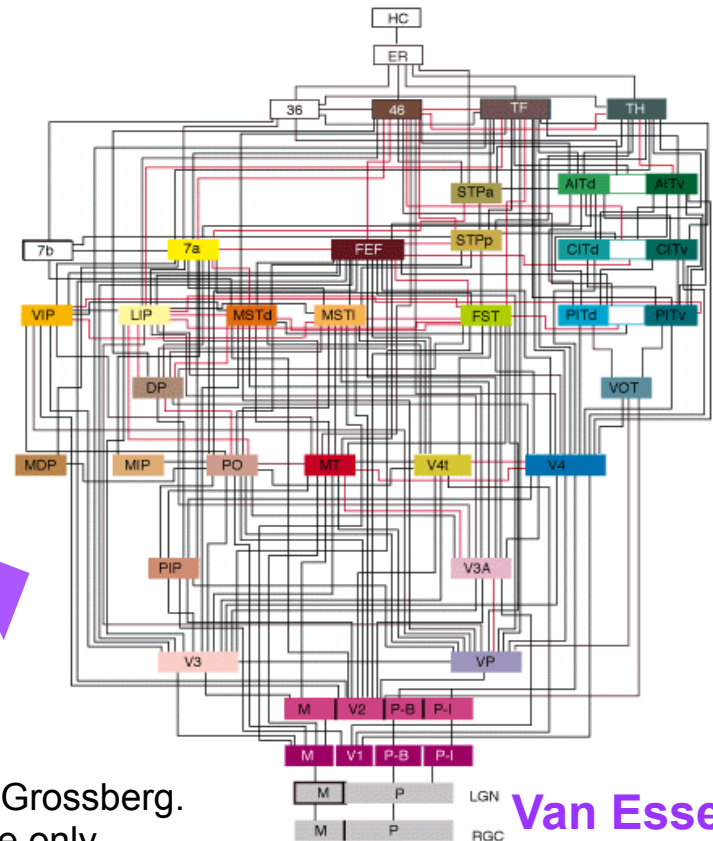
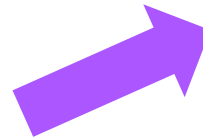
COMPLEMENTARY COMPUTING

New principles of
UNCERTAINTY and **COMPLEMENTARITY**
clarify why

Multiple Parallel Processing Streams Exist in the Brain



Lots of specialization!



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Van Essen et al

WHAT ARE COMPLEMENTARY PROPERTIES?

Analogies:

Key fitting in lock, puzzle pieces fitting together



Computing one set of properties at a processing stage prevents that stage from computing a **complementary** set of properties

Complementary parallel processing streams are **BALANCED** against one another

INTERACTIONS between streams overcomes their complementary weaknesses and support intelligent and creative behaviors

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SOME COMPLEMENTARY PROCESSES

Visual Boundary

Interbob Stream V1-V4

Visual Surface

Blob Stream V1-V4

Visual Boundary

Interbob Stream V1-V4

Visual Motion

Magno Stream V1-MT

WHAT Steam

Perception & Recognition

Inferotemporal and

Prefrontal areas

WHERE Stream

Space & Action

Parietal and

Prefrontal areas

Object Tracking

MT Interbands and MSTv

Optic Flow Navigation

MT Bands and MSTd

Motor Target Position

Motor and Parietal Cortex

Volitional Speed

Basal Ganglia

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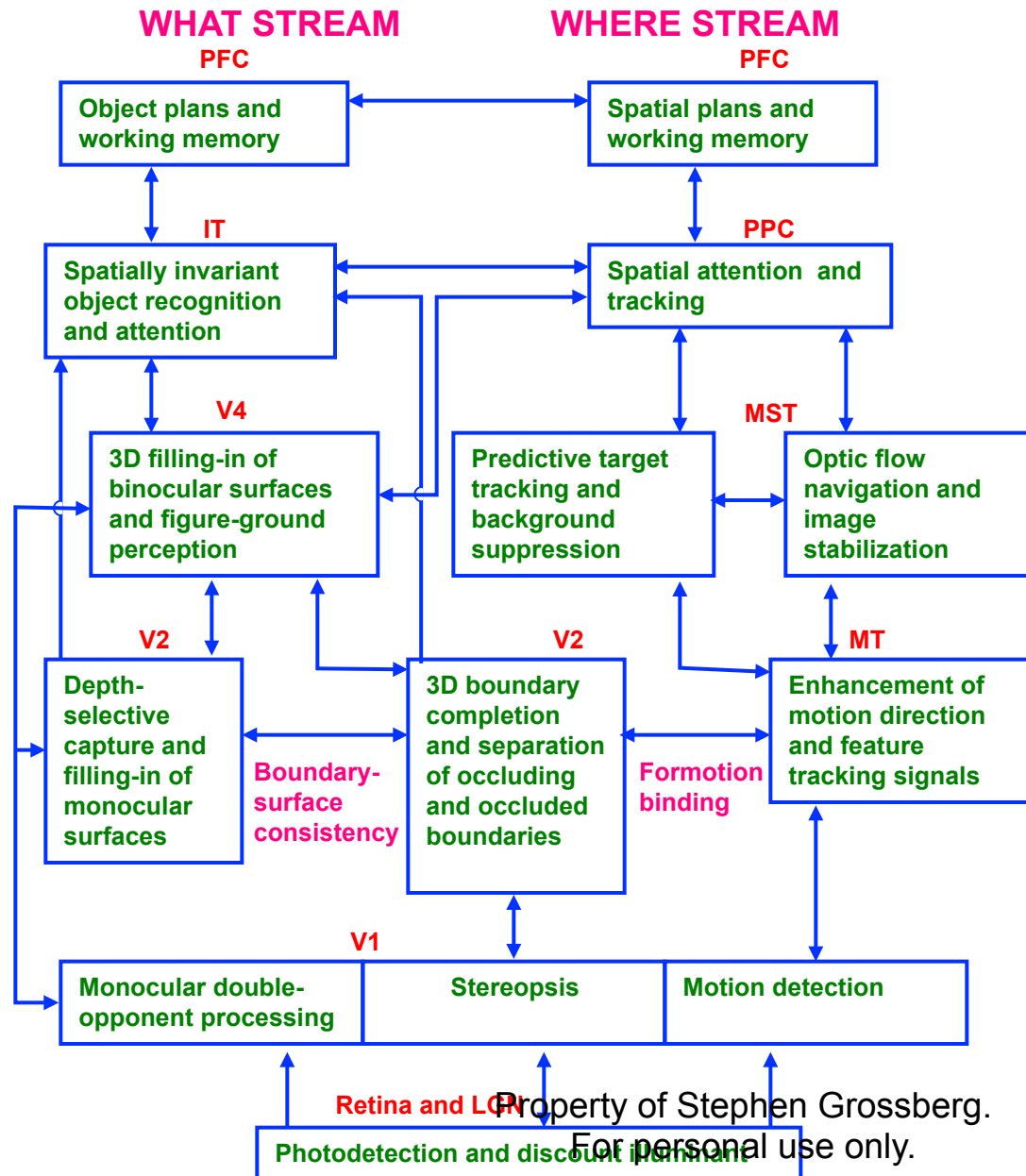
Object Tracking
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Optic Flow Navigation
MT Bands and MSTd

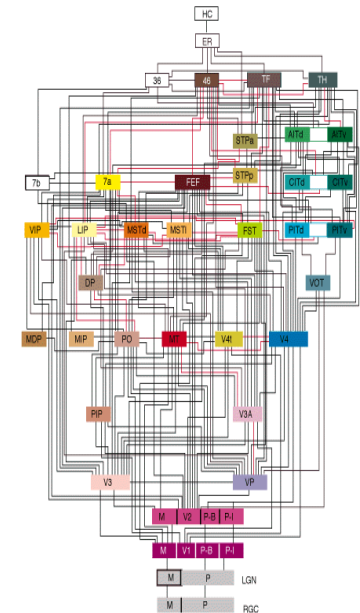
Motor Target Position
Motor and Parietal Cortex

Volitional Speed
Basal Ganglia

EMERGING UNIFIED THEORY OF VISUAL INTELLIGENCE



Bottom-up
horizontal and
top-down
interactions
overcome
COMPLEMENTARY
processing
deficiencies



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SOME COMPLEMENTARY PROCESSES

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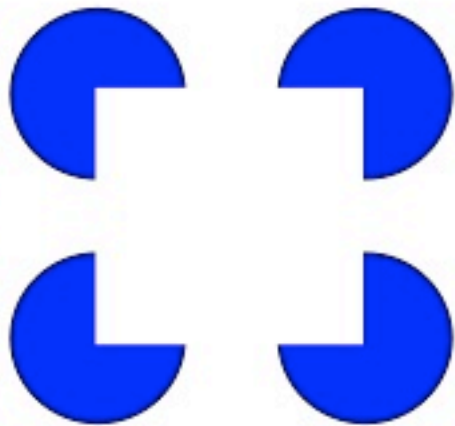
Optic Flow Navigation
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Motor Target Position
Motor and Parietal Cortex

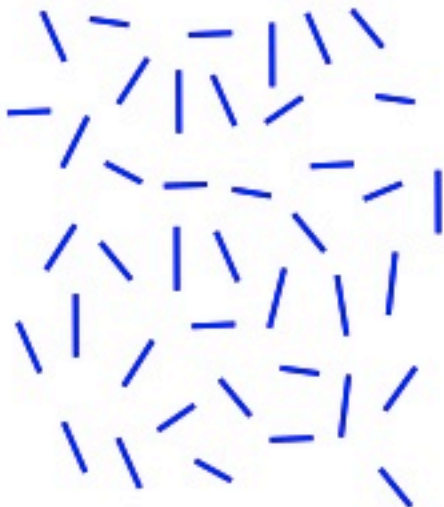
Volitional Speed
Basal Ganglia

WHAT IS A VISUAL BOUNDARY OR GROUPING?

Illusory
contour



Texture
pop-out



3D shape
from
texture

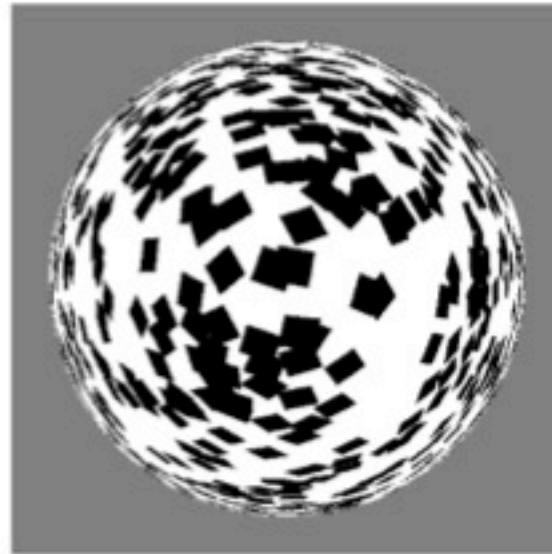
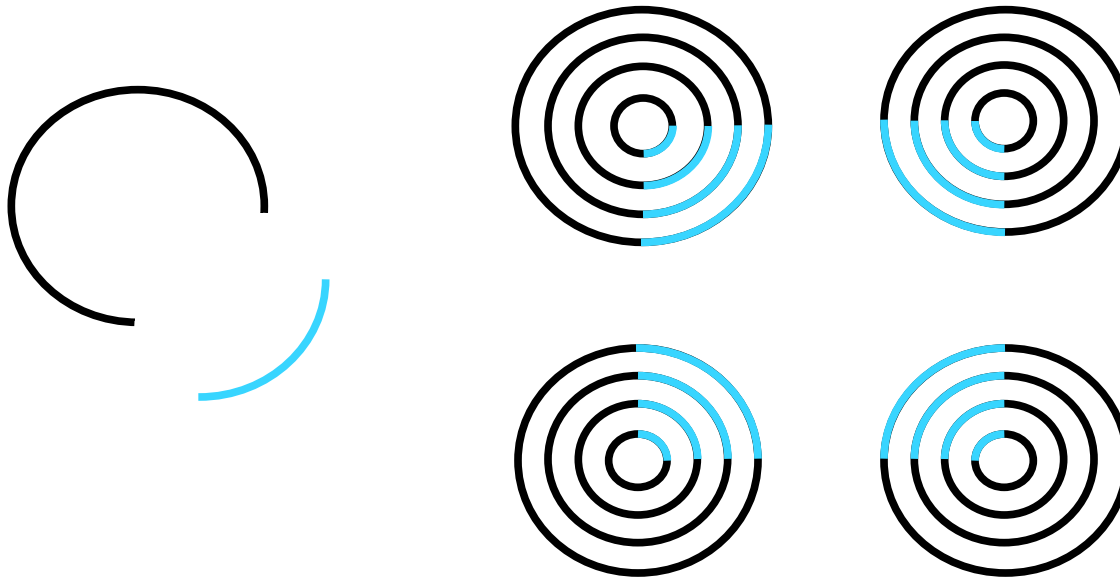


Figure-
ground
separation



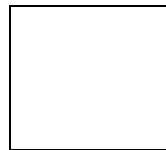
VISUAL BOUNDARY AND SURFACE COMPUTATIONS ARE COMPLEMENTARY

Grossberg (1984)



Neon color spreading

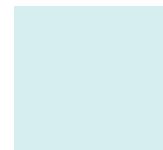
**BOUNDARY
COMPLETION**



oriented
inward

insensitive to
direction-of-contrast

**SURFACE
FILLING-IN**



unoriented
outward

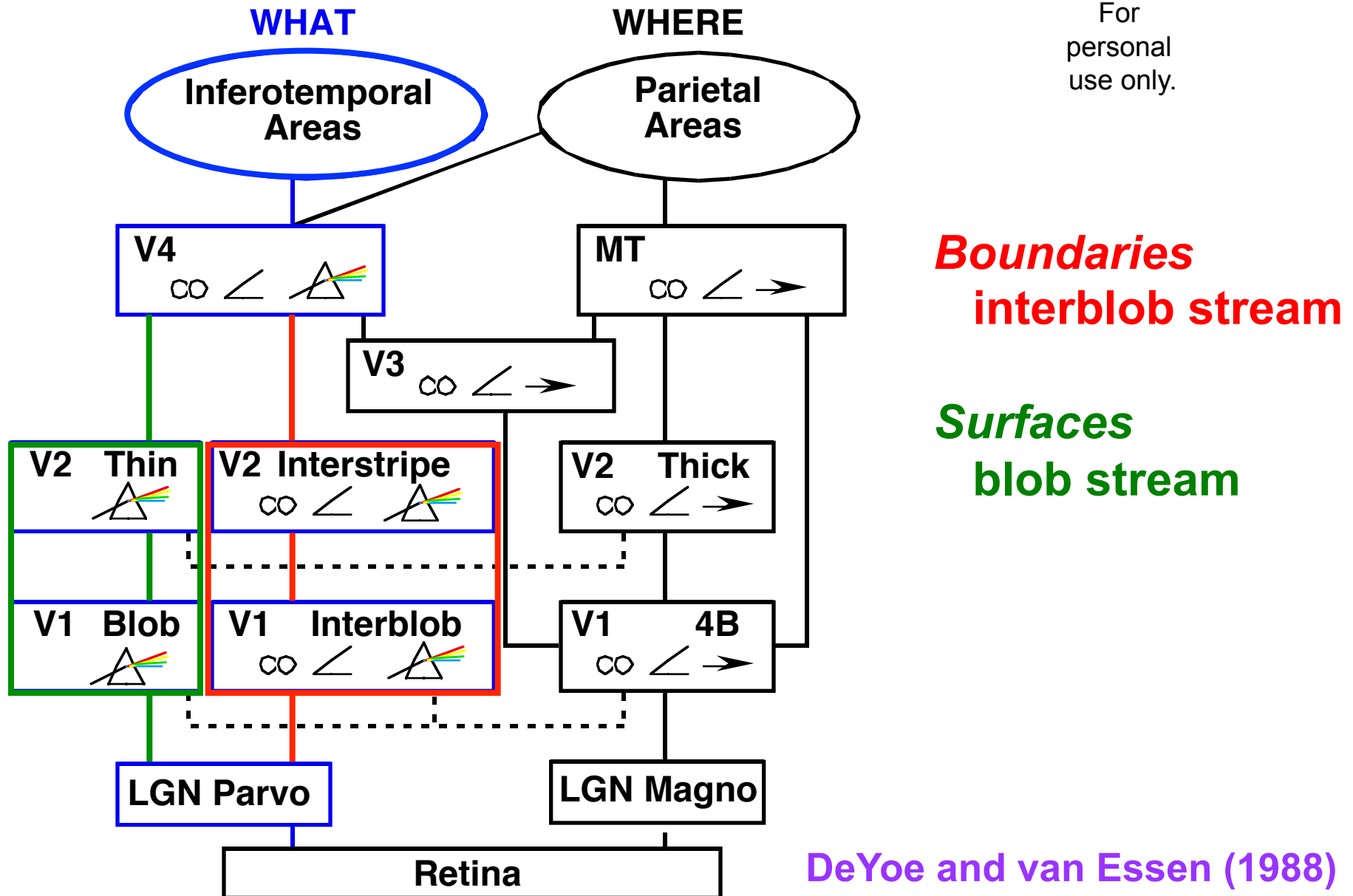
sensitive to
direction-of-contrast

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BOUNDARY AND SURFACE CORTICAL STREAMS

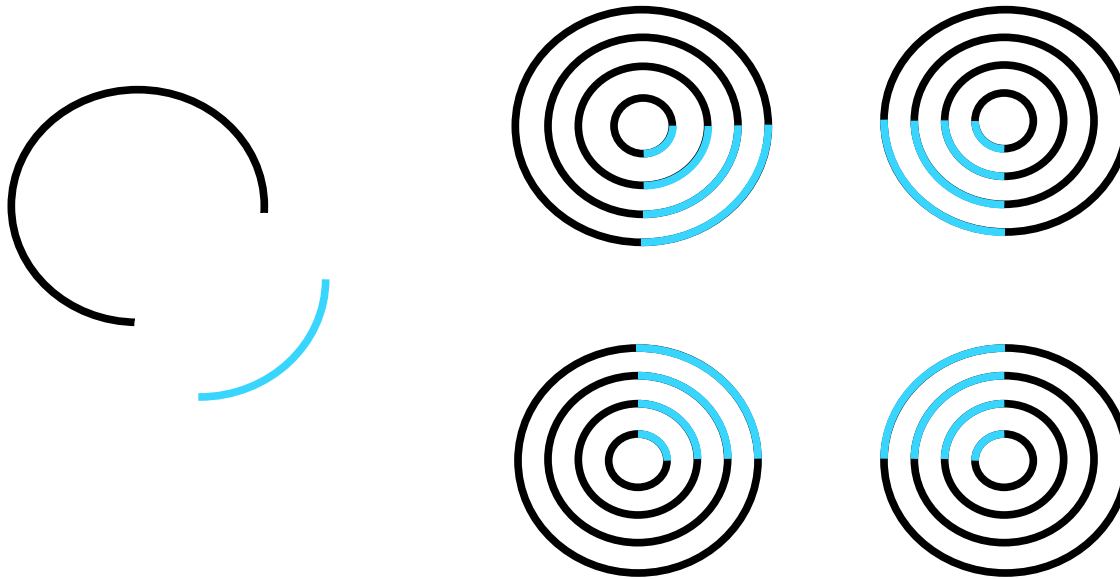
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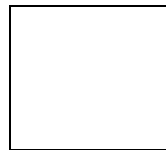
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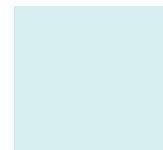
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What about



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SEEING vs. KNOWING

SEEING
an object

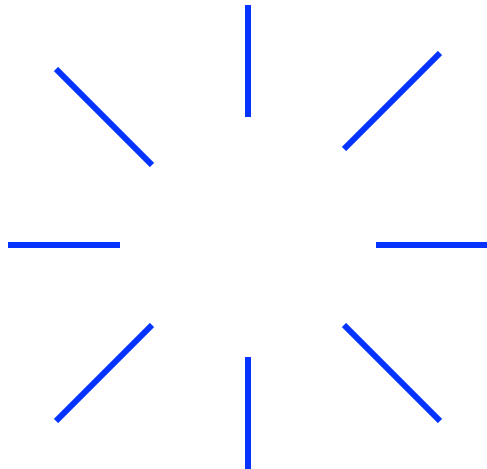
vs.

KNOWING
what it is

Epstein, Gregory, Helmholtz, Kanizsa, Kellman, Michotte,...

SEEING

Ehrenstein Figure

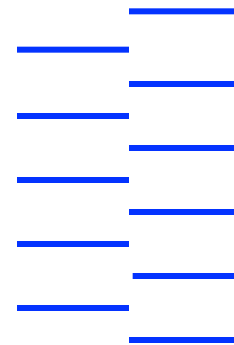


See
Recognize

vs.

RECOGNIZING

Offset Grating



Some
boundaries
are
invisible,
or amodal

Do not see
Recognize

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ALL BOUNDARIES ARE INVISIBLE!

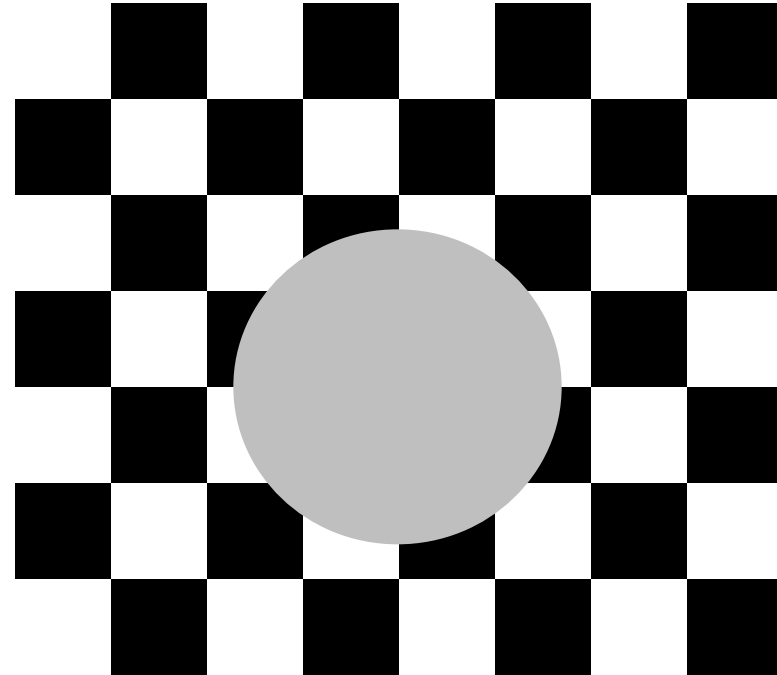
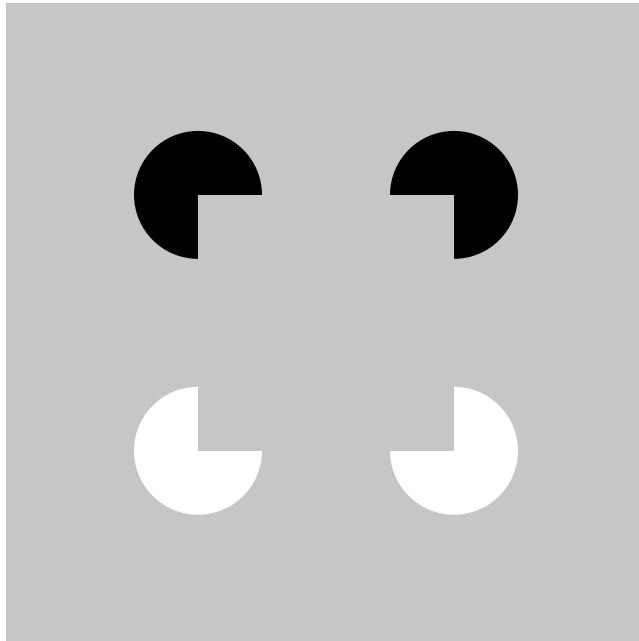
Within the Boundary Steam

Grossberg (1984)

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30

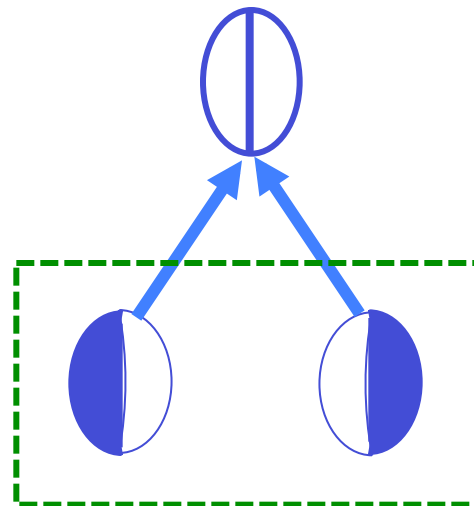
WHY? To recognize object boundaries in front of textured backgrounds



ALL BOUNDARIES ARE INVISIBLE: COMPLEX CELLS

complex cells pool inputs from
opposite-polarity simple cells in V1

V1



complex cells

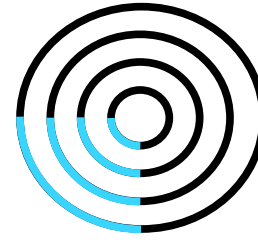
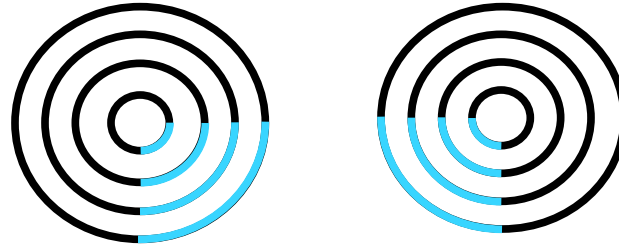
simple cells

Complex cells are amodal boundary detectors Grossberg (1984)

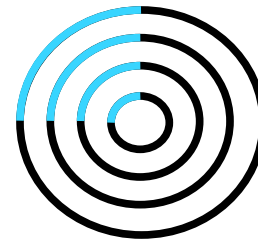
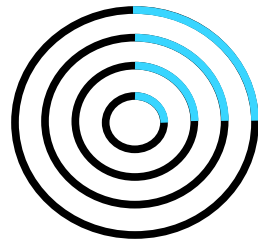
vs

“color cells in the broadest sense” Thorell, DeValois & Albrecht (1984)

VISUAL BOUNDARY AND SURFACE COMPUTATIONS ARE COMPLEMENTARY



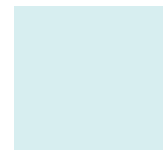
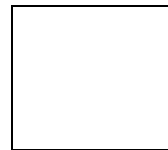
Neon color spreading



All Boundaries
Are
Invisible!

**BOUNDARY
COMPLETION**

**SURFACE
FILLING-IN**



**oriented
inward**

**unoriented
outward**

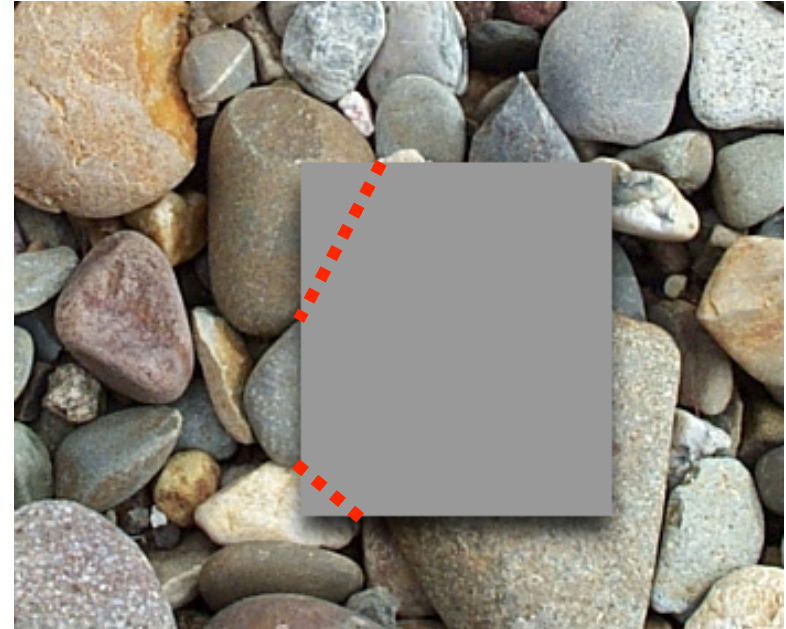
insensitive to direction-of-contrast **sensitive to direction-of-contrast**

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MANY OBJECTS ARE PARTIALLY OCCLUDED IN A 3-DIMENSIONAL WORLD



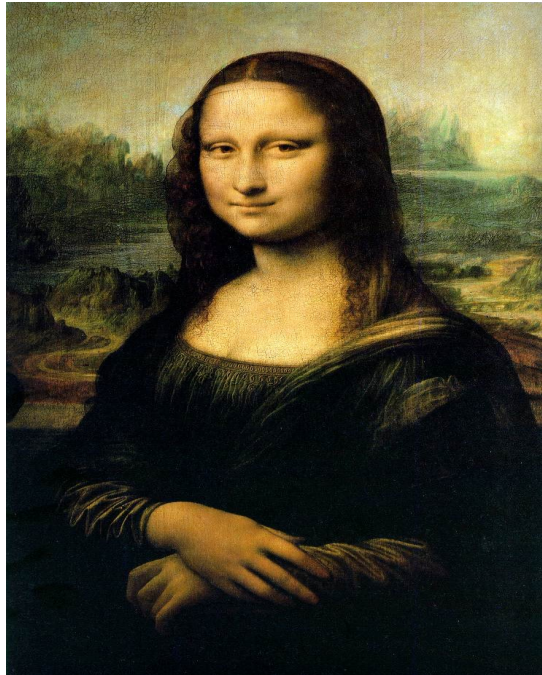
MANY OBJECTS ARE PARTIALLY OCCLUDED IN A 3-DIMENSIONAL WORLD



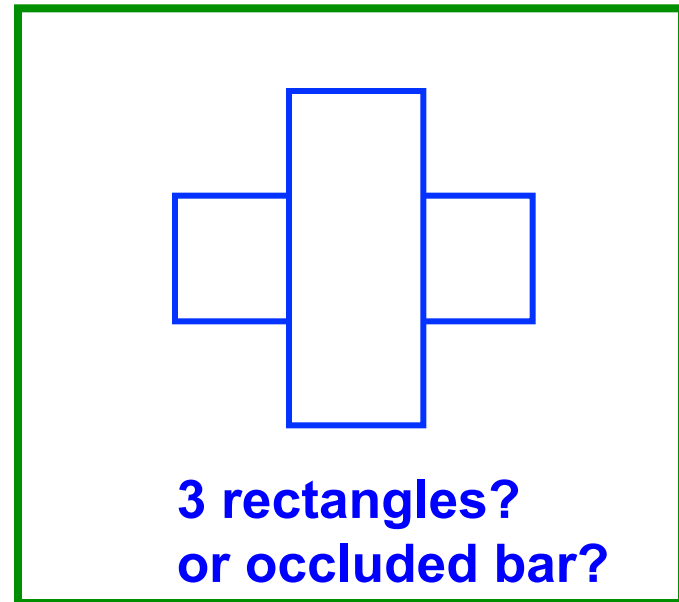
**Amodal, or invisible, completion of partially occluded objects
behind their occluding objects allows us to better recognize them**

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MANY OBJECTS ARE PARTIALLY OCCLUDED IN A 3-DIMENSIONAL WORLD



Mona Lisa

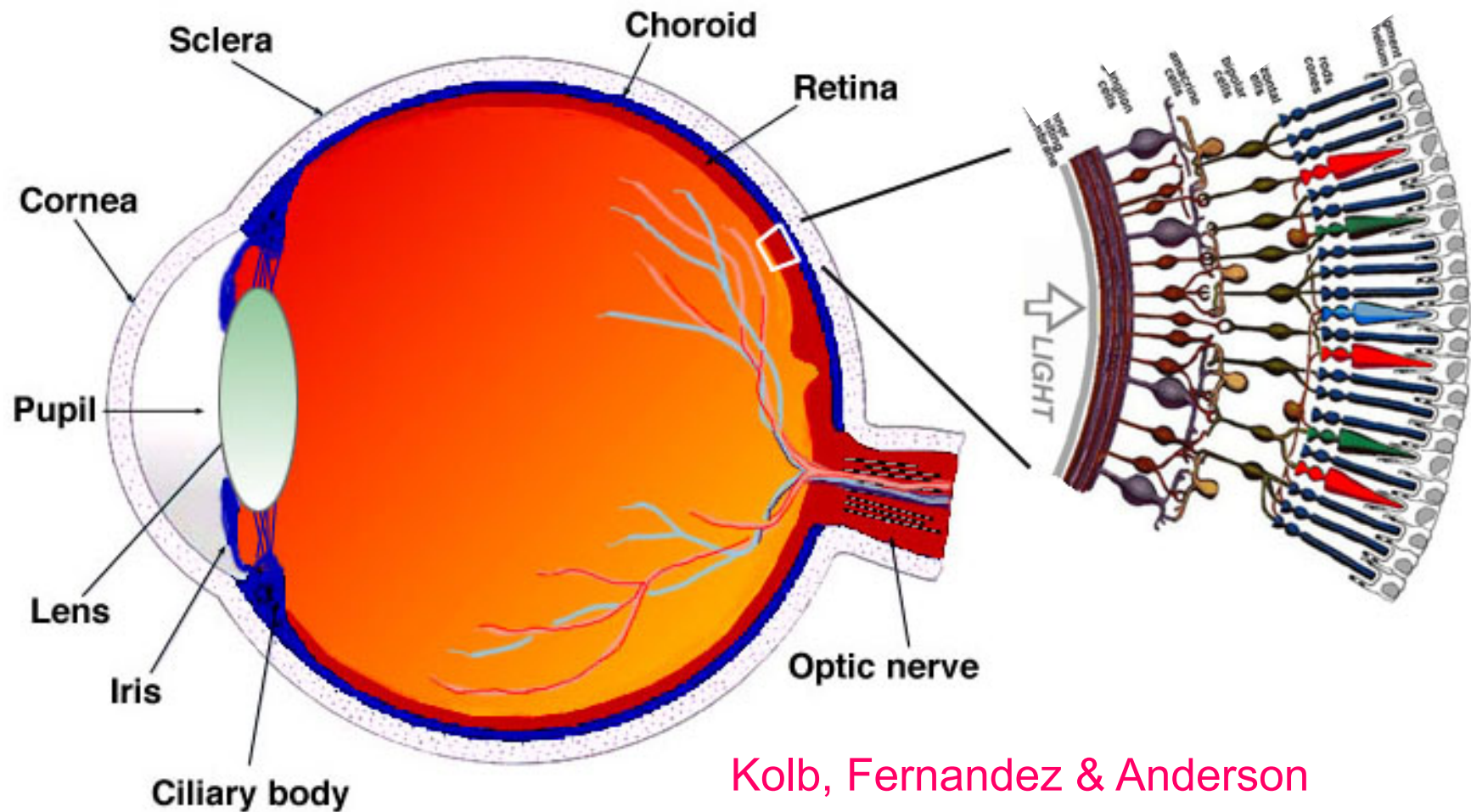


This same process allows 2D pictures to be perceived as 3D representations of occluding and occluded surfaces

It is a key process in all pictorial art, movies, and TV

BLIND SPOT AND RETINAL VEINS

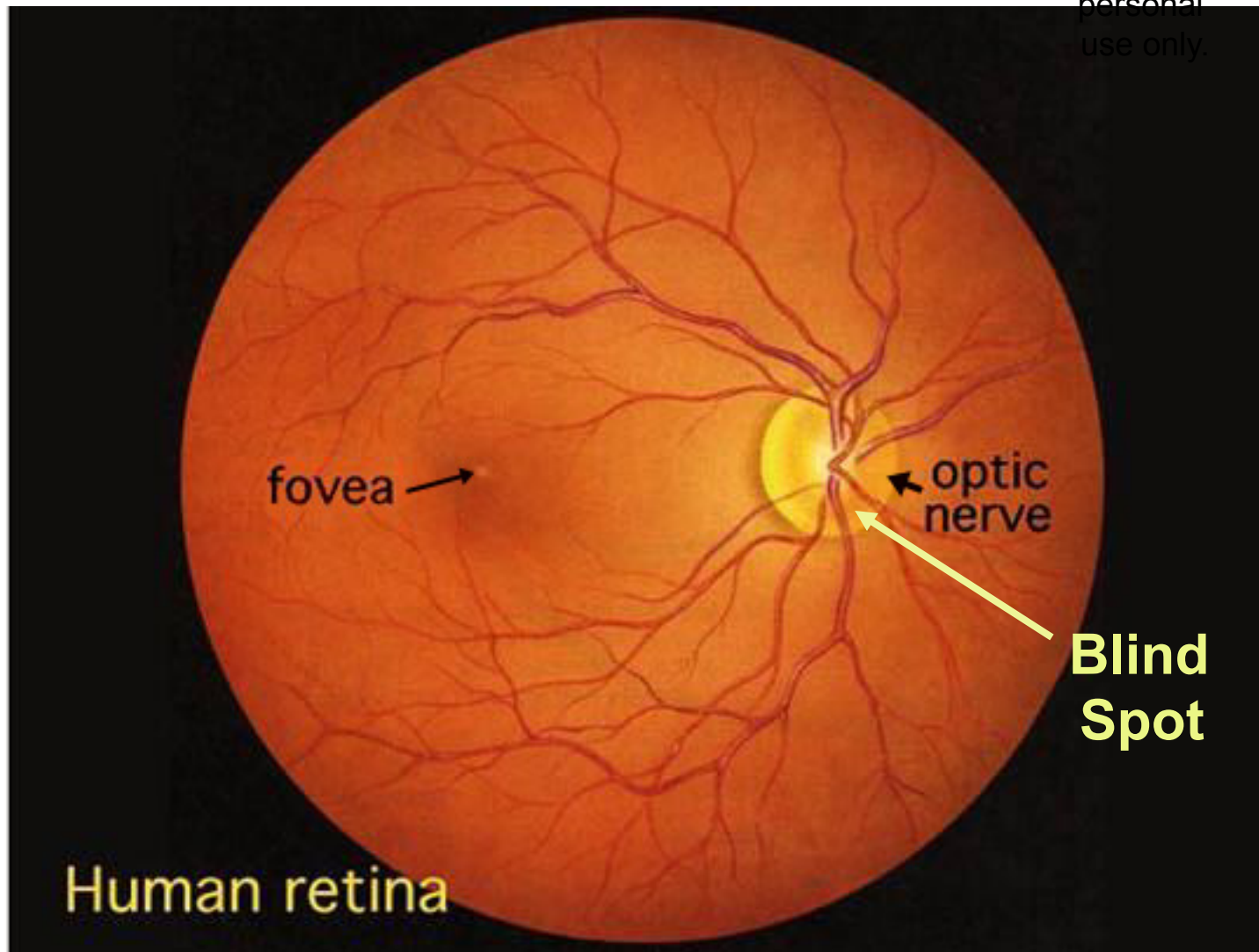
another reason for boundary completion and surface filling-in



Kolb, Fernandez & Anderson

<http://retina.umh.es/Webvision/sretina.html>

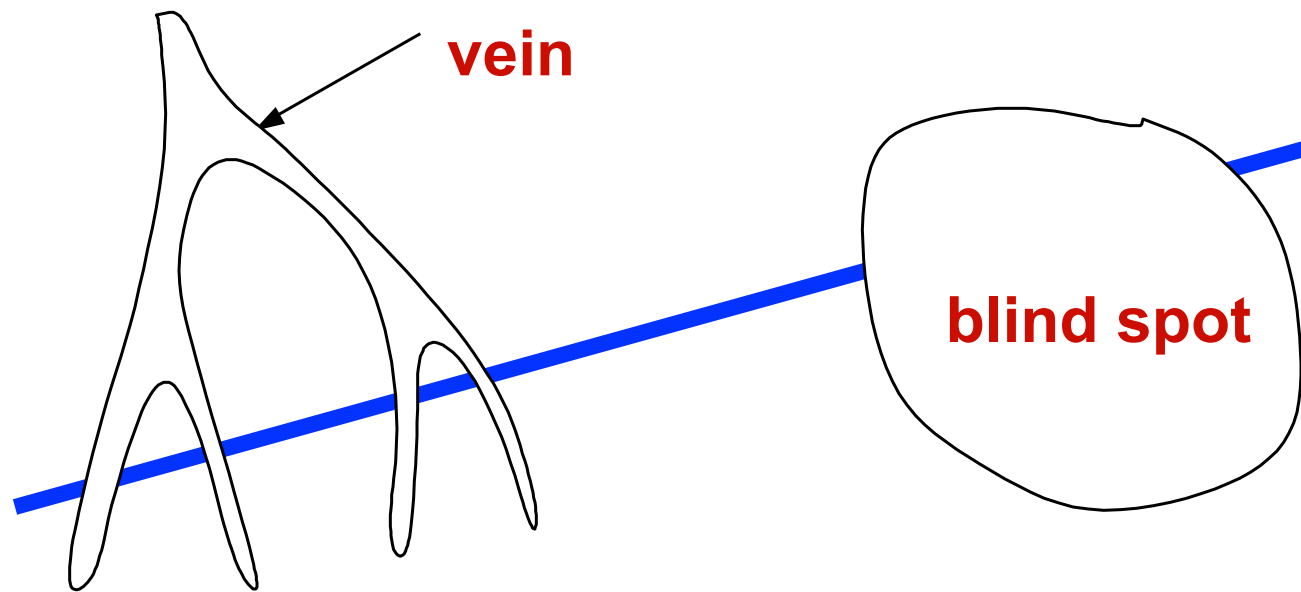
TOP-DOWN VIEW OF THE RETINA



Blind spot, retinal veins, and layers all interfere

WHY DON'T WE SEE BLIND SPOT AND RETINAL VEINS?!

The pattern formed on a retina by a dark line



...is not even connected!

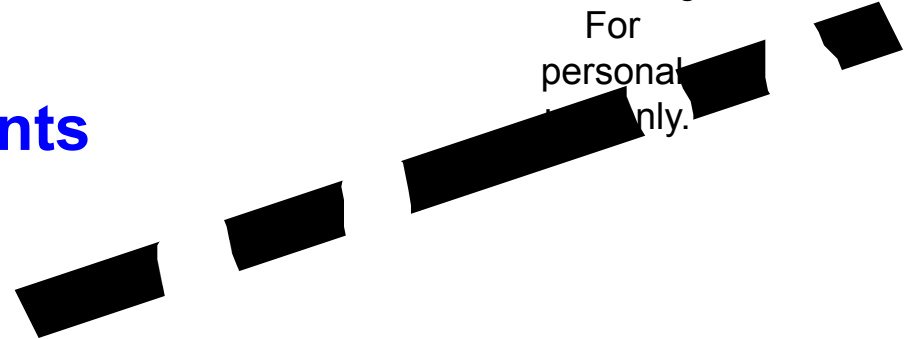
Eye jiggles in its orbit
Stabilized images fade

EVERY LINE IS AN ILLUSION!

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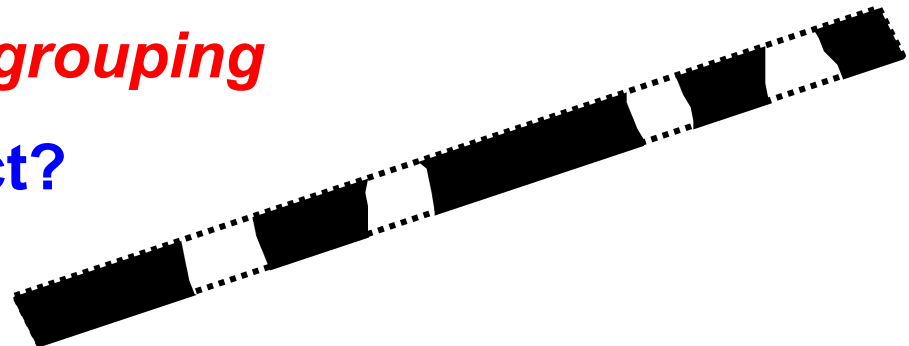
39

Line is registered as fragments



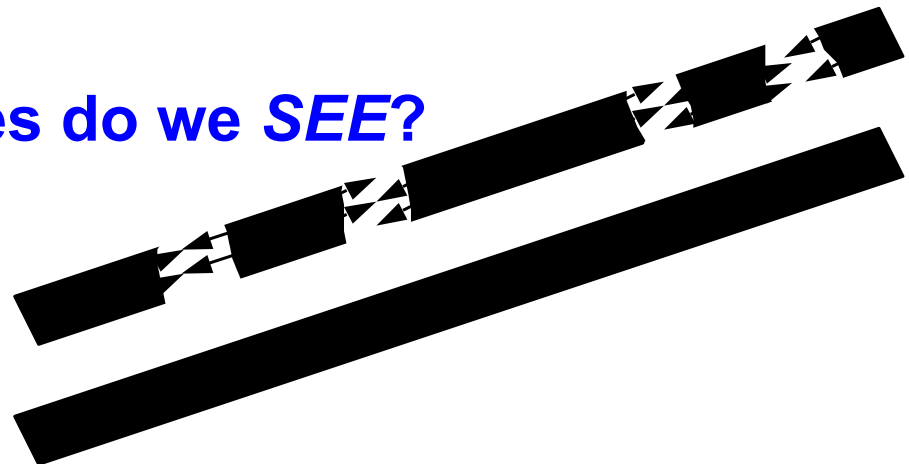
Boundary completion and grouping

Which boundaries to connect?



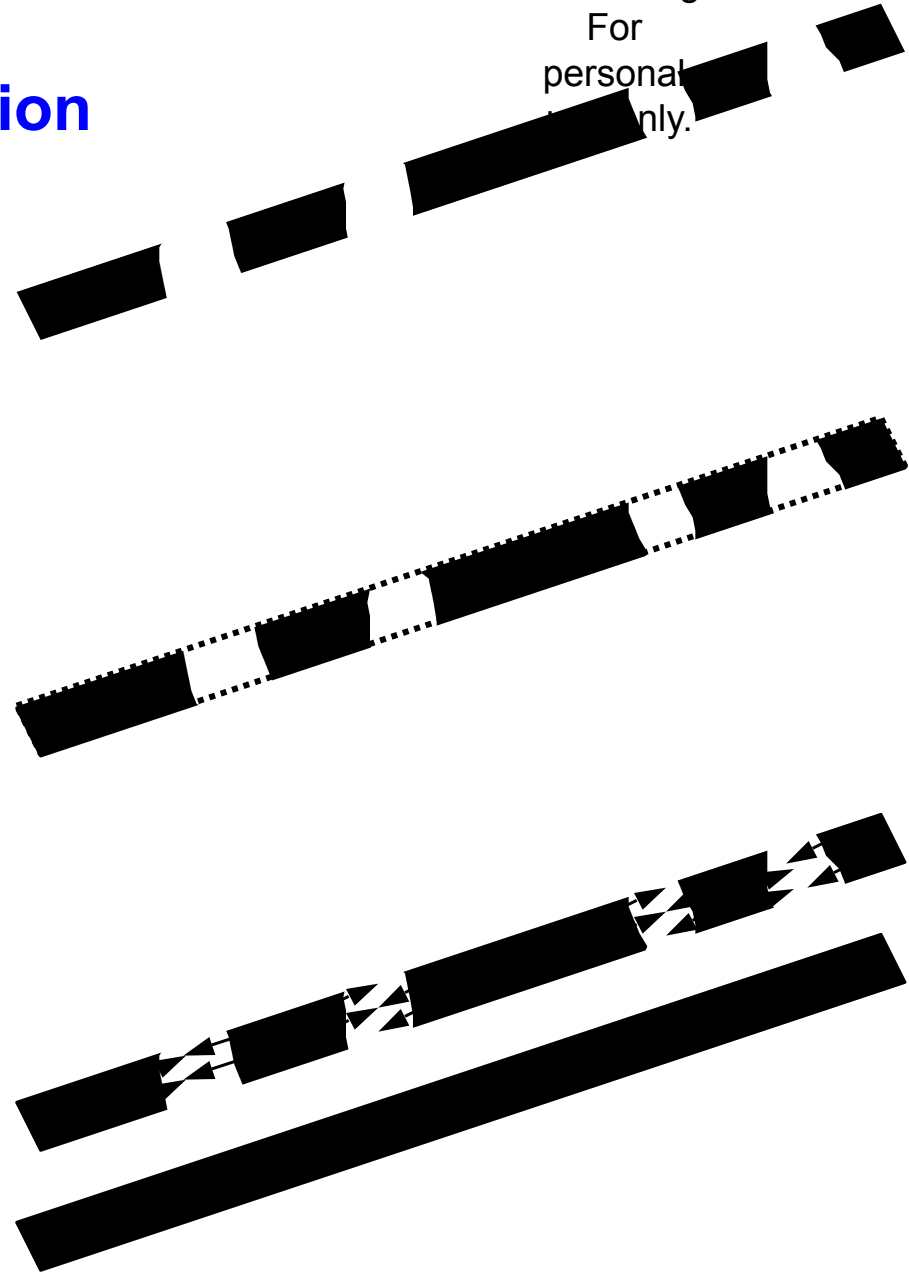
Surface filling-in

What colors and brightnesses do we SEE?



WHAT DO WE CALL AN ILLUSION?

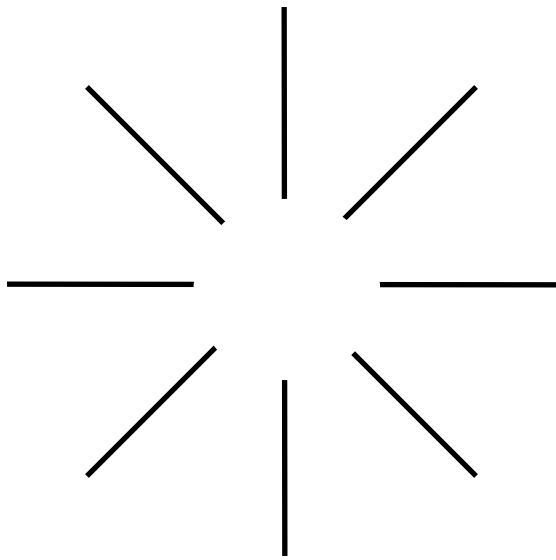
...an **unexpected** combination
of boundary completion
and surface filling-in



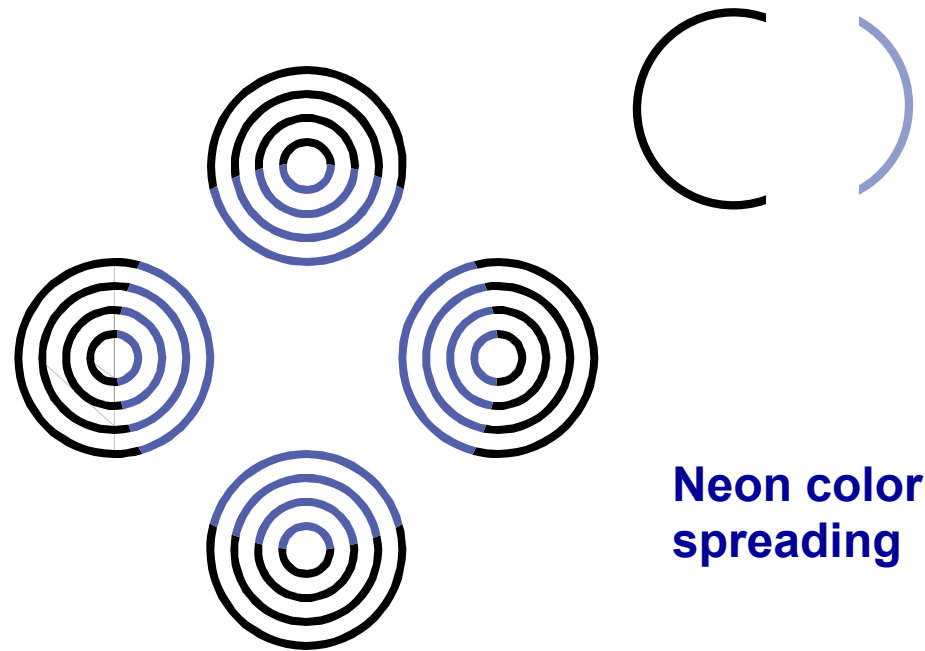
IF BOUNDARIES ARE INVISIBLE, HOW DO WE SEE?

Filling-In of Surface Color

Boundaries define the compartments
within which lightness and color spread



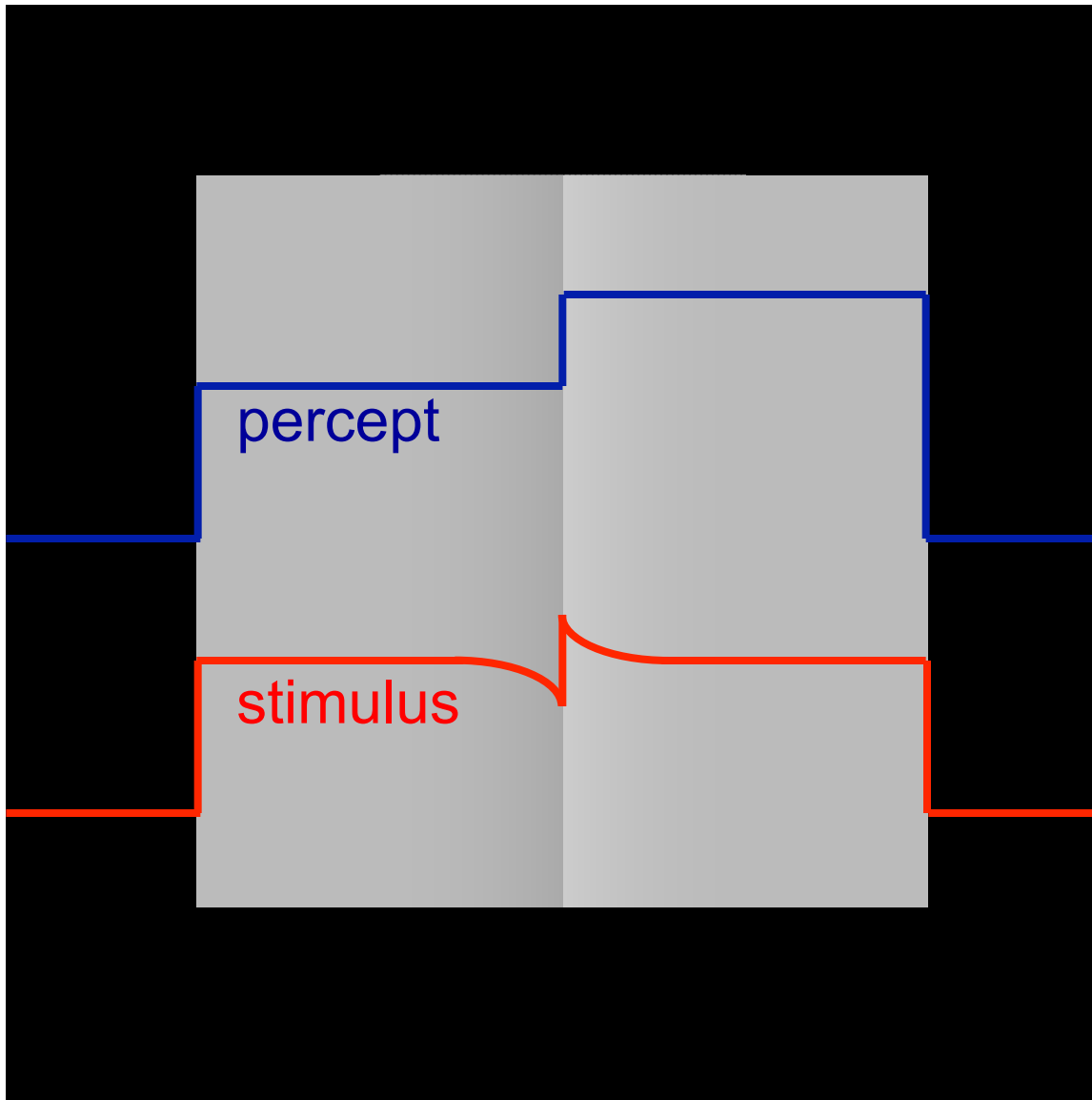
Ehrenstein (1941)



Varin (1971)

Neon color
spreading

Craik-O'Brien-Cornsweet Effect

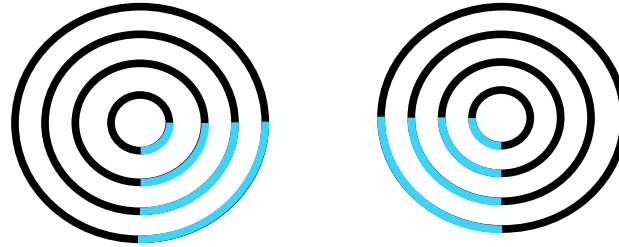


**Boundary completion
defines
filling-in compartments**

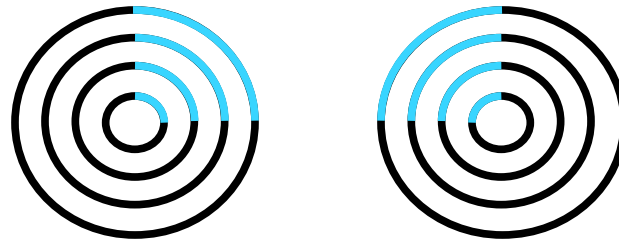
**Filling-in determines
what we see
in each compartment**

Grossberg (1984)
Todorović (1987)

VISUAL BOUNDARY AND SURFACE COMPUTATIONS ARE COMPLEMENTARY



Neon color spreading

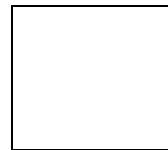


All Boundaries
Are
Invisible!

Filling-in of
Visible
Color and
Lightness

BOUNDARY
COMPLETION

SURFACE
FILLING-IN



oriented
inward

unoriented
outward

insensitive to direction-of-contrast

sensitive to direction-of-contrast

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PREDICTIONS

Grossberg (1984)

ALL BOUNDARIES ARE INVISIBLE
in the interblob stream

VISIBLE QUALIA ARE SURFACE PERCEPTS
in the blob stream

WHAT IS THIS TALK ABOUT?

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Jo Baer, Banksy, Ross Bleckner, Gene Davis, Charles Hawthorne, Henry Hensche, Henri Matisse, Claude Monet, Jules Olitski, Frank Stella

DID ARTISTS LIKE MATISSE KNOW THAT ALL BOUNDARIES ARE INVISIBLE?!

Yes: It was the basis of the Fauve artistic movement



Matisse, The Roofs of Collioure, 1905

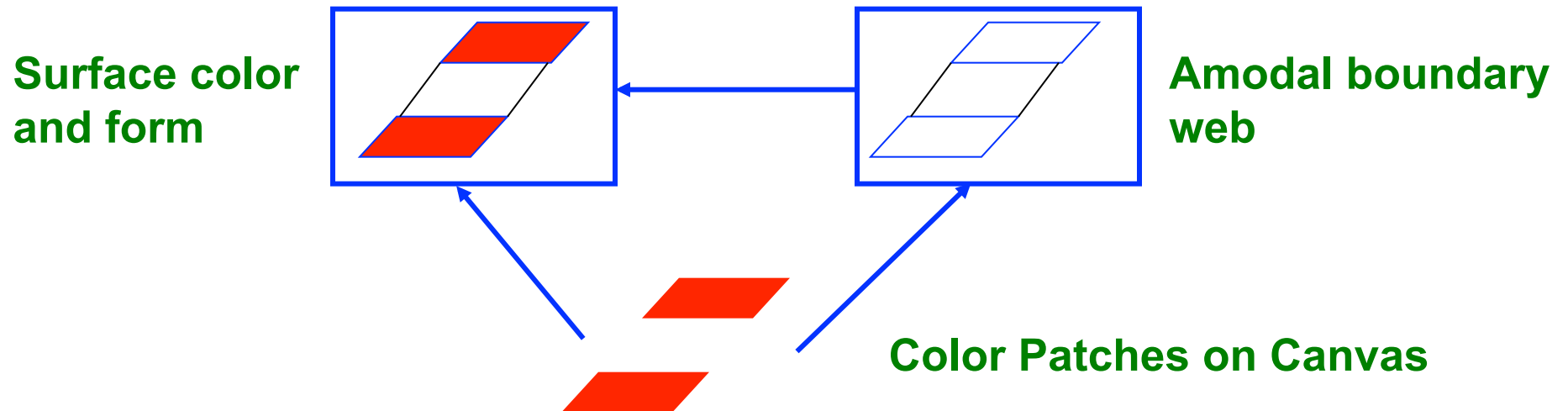
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MATISSE KNEW THAT ALL BOUNDARIES ARE INVISIBLE

He went through a life-long struggle to understand
“the eternal conflict between drawing and color”

“Instead of drawing an outline and filling in the color...
I am drawing directly in color”

H. Matisse, Jazz (1947)



Drawing an explicit boundary could **DARKEN THE COLORS!**
via color assimilation

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COMPLEMENTARITY! MANY INVISIBLE BOUNDARIES!



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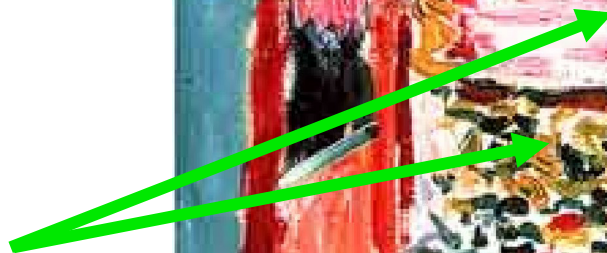
Matisse, The Roofs of Collioure, 1905

Continuously
induced
surface



Matisse,
Open
Window,
Collioure
1905

Sparsely
induced
surfaces



Property of Stephen Grossberg.
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DRAWING DIRECTLY IN COLOR

**Many artists have experienced Matisse's struggle to be
“DRAWING DIRECTLY IN COLOR”**

PLEIN AIR PAINTERS

CAPE COD SCHOOL OF ART: DRAWING IN COLOR

Charles Hawthorne:

“Beauty in art is the delicious notes of color one against the other ...all we have to do is to get the color notes in their proper relation... put down spots of color...the outline and size of each spot of color against every other spot of color it touches, is the only kind of drawing you need bother about...**Let color make form—do not make form and color it. Forget about drawing...**drawing the form, and painting, are better separated. The first thing is to learn to see color...”

Hawthorne, C. W. (1938/1960). *Hawthorne on painting*. Mineola, New York: Dover.

PLEIN AIR PAINTERS

CAPE COD SCHOOL OF ART: DRAWING IN COLOR

Henry Hensche:

When Monet came along...he revolutionized the 'art of seeing.'
...The landscape helped Monet determine how color expressing
the light key was the first ingredient in a painting, not drawing...
Every form change must be a color change..."

Robichaux, J. W. (1997). *Hensche on painting*. Mineola, New York: Dover.

PLEIN AIR PAINTERS

FRENCH IMPRESSIONISTS: DRAWING IN COLOR

Claude Monet:

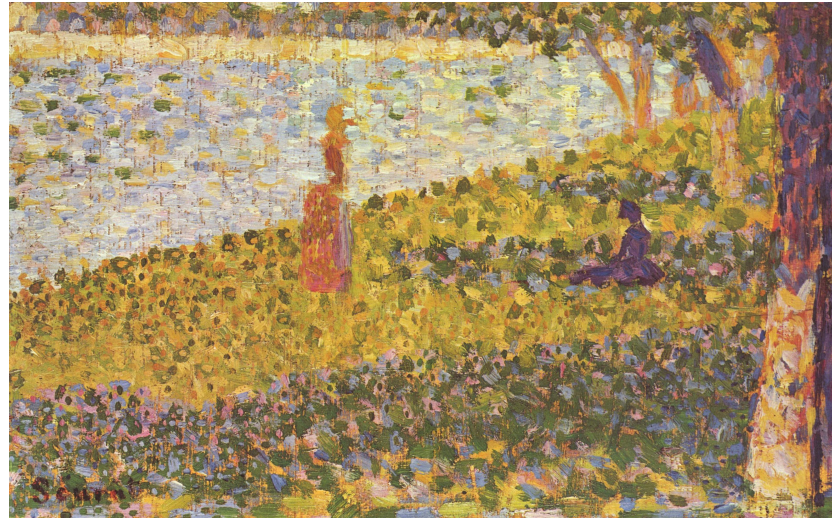
“When you go out to paint, try to forget what objects you have before you, a tree, a house, a field, or whatever. Merely think, here is a little square of blue, here an oblong of pink ... paint it just as it looks to you, the exact color and shape, until it gives your own naïve impression of the scene before you...”

Perry, 1927, p. 120

FRENCH IMPRESSIONISTS : SEURAT

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54



Seurat, Femmes au bord de l' eau

Boundaries complete between regions where feature contrasts change

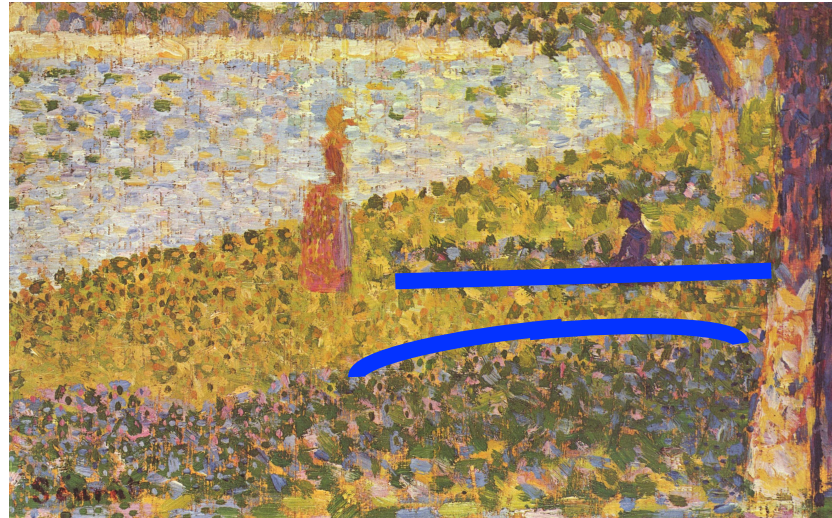
**Boundary groupings organize filling-in of colors within them
to form visible surface percepts**

FRENCH IMPRESSIONISTS : SEURAT

On a large scale

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55



Seurat, Femmes au bord de l' eau

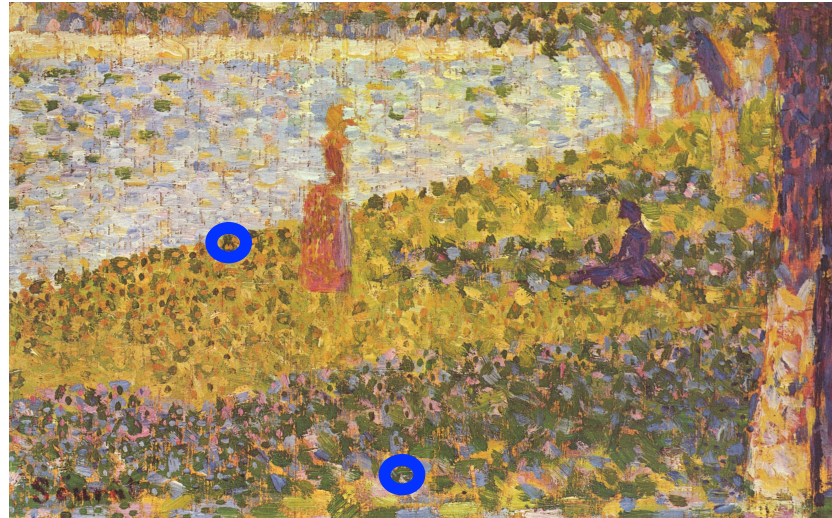
Boundaries complete between regions where feature contrasts change

**Boundary groupings organize filling-in of colors within them
to form visible surface percepts**

FRENCH IMPRESSIONISTS : SEURAT

And a small scale...e.g., individual dots

Property of
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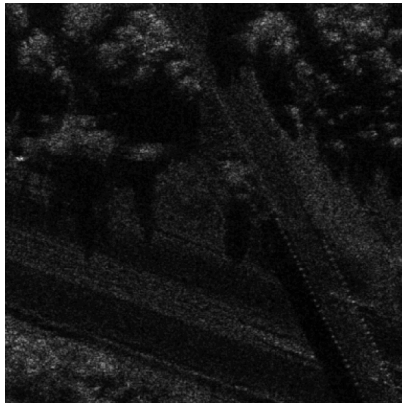
Seurat, Femmes au bord de l' eau

Boundaries complete between regions where feature contrasts change

Boundary groupings organize filling-in of colors within them
to form visible surface percepts

DO THESE IDEAS WORK ON HARD PROBLEMS?

From Seurat to SAR



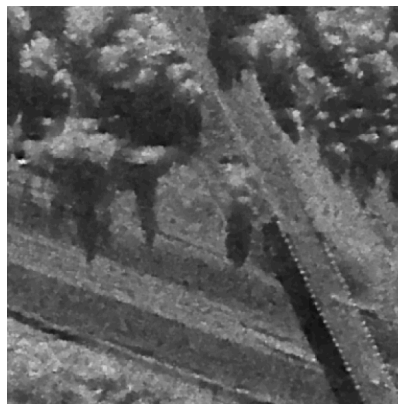
input



feature



boundary



surface filling-in

Application: Image Enhancement

Synthetic aperture radar
sees through the weather

signal: 5 orders of magnitude
of power in radar return

multiplicative noise

sparse high-intensity pixels

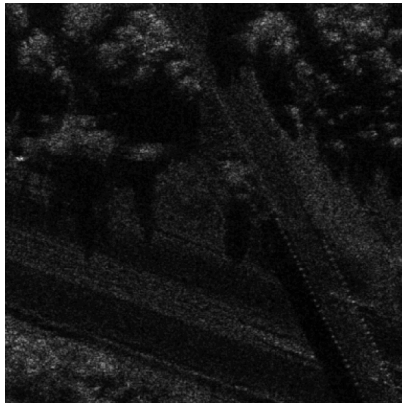
Clarifies how we see
an Impressionistic
painting

Mingolla, Ross, and Grossberg (1989)

Property of Stephen Grossberg. For
personal use only.

DO THESE IDEAS WORK ON HARD PROBLEMS?

From Seurat to SAR



input



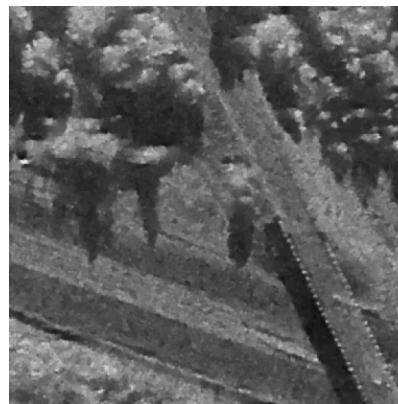
feature

Discounting the Illuminant
normalizes the image:
It preserves **RELATIVE** activities
without **SATURATION**

Still shows individual **PIXELS**



boundary



surface filling-in

Filling-in averages brightnesses
within boundary compartments

Boundaries complete between
regions where normalized
feature contrasts change



Property of Stephen Grossberg. For
personal use only.

MULTIPLE SCALE BOUNDARIES AND FILLING-IN

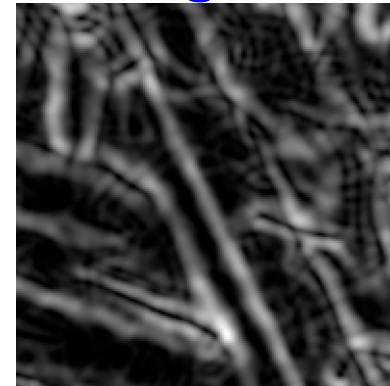
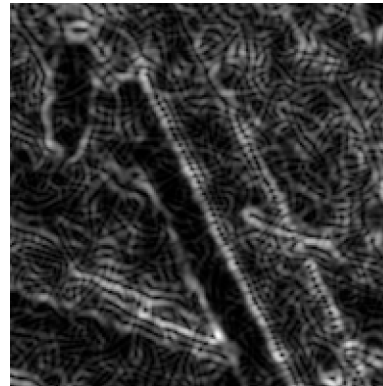
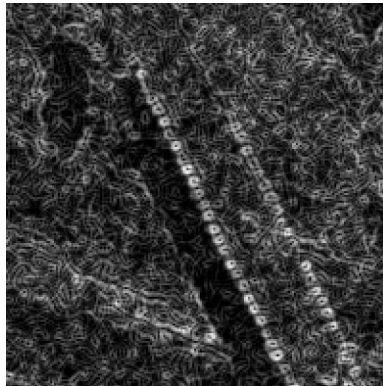
Scale:

small

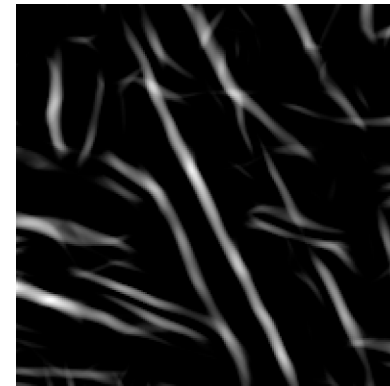
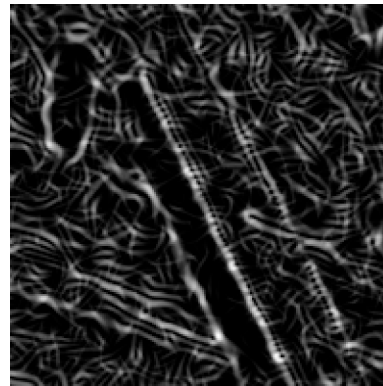
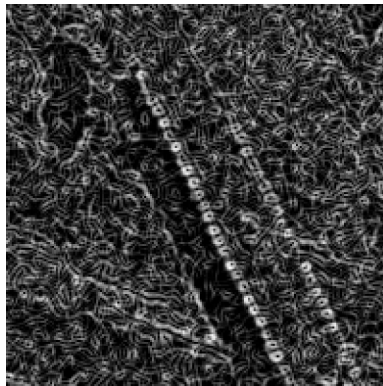
medium

large

boundaries
before
completion



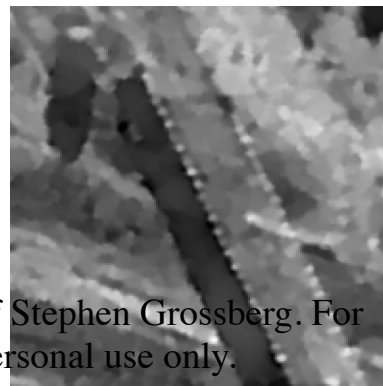
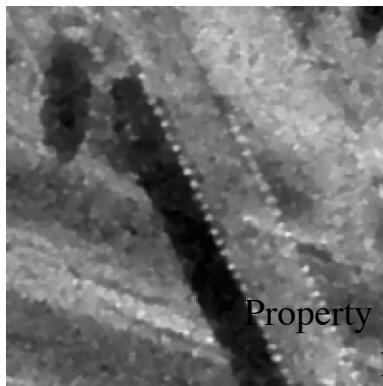
boundaries
after
completion



*large
scale
bipole:*



surface
filling-in

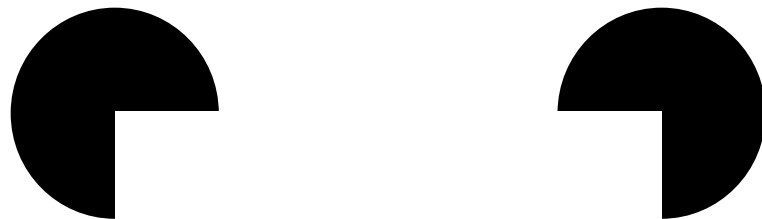


Property of Stephen Grossberg. For
personal use only.

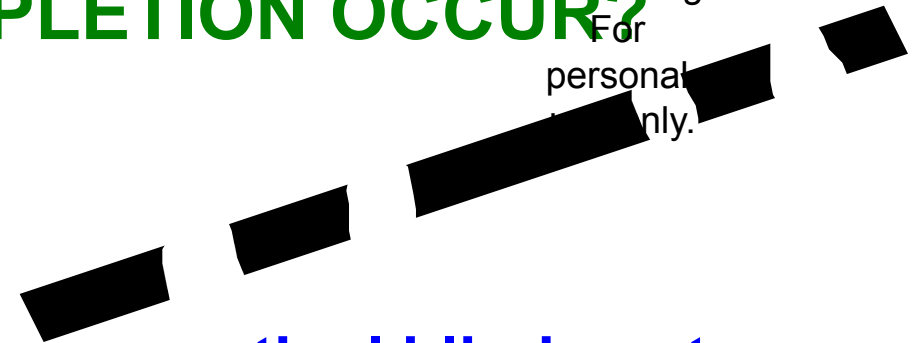
HOW DOES INWARD, ORIENTED BOUNDARY COMPLETION OCCUR?

Property of
Stephen
Grossberg.
For
personal
use only.

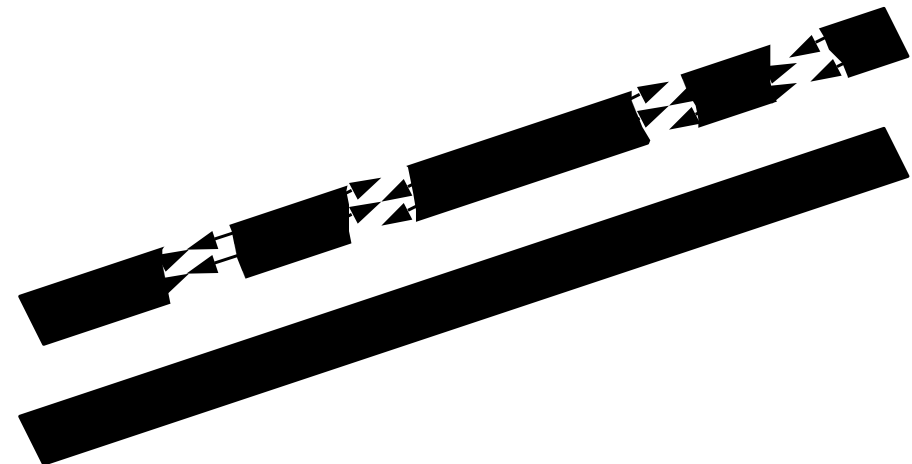
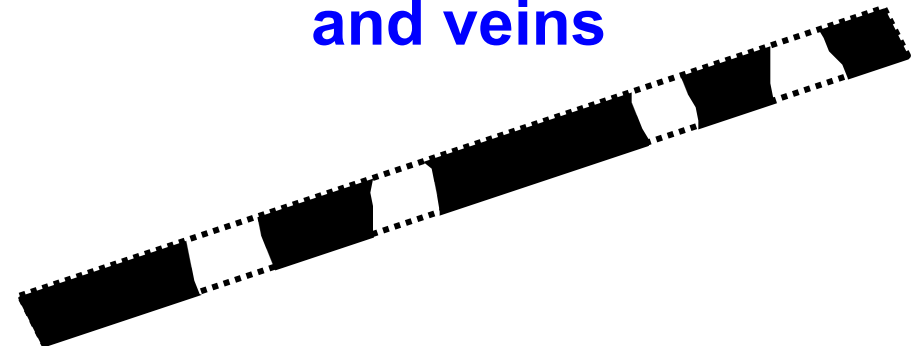
60



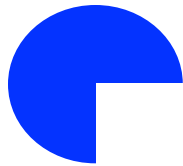
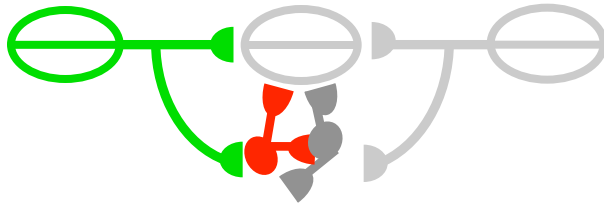
Kanizsa square



retinal blind spot
and veins

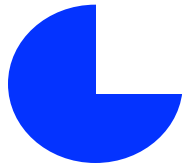


1984 PREDICTION: BIPOLE PROPERTY CONTROLS PERCEPTUAL GROUPING



Input on just **one side**

ONE-AGAINST-ONE:
Balanced **Excitation** and **Inhibition**



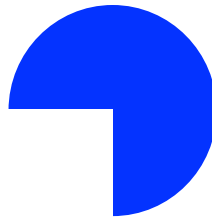
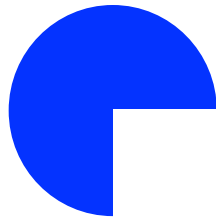
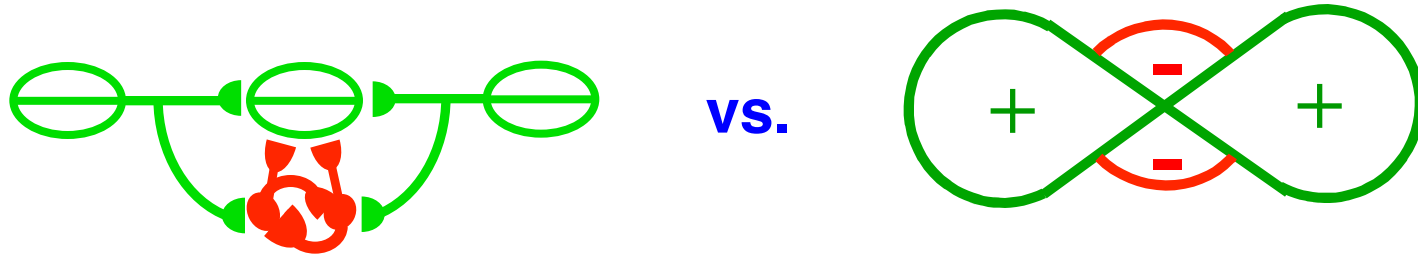
Cell **not** excited

Grossberg, 1984

Grossberg & Mingolla, 1985

Laminar: Grossberg, Mingolla & Ross, 1997

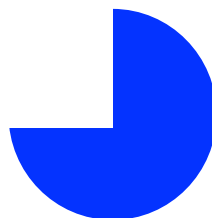
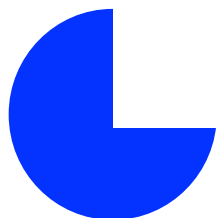
BIPOLE PROPERTY CONTROLS PERCEPTUAL GROUPING



Collinear input on **both sides**

Excitatory inputs summate

Inhibitory inputs normalize
Shunting inhibition!









TWO-AGAINST-ONE

Cell is excited

BIPOLES: FIRST NEUROPHYSIOLOGICAL EVIDENCE from cortical area V2

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personal
use only.

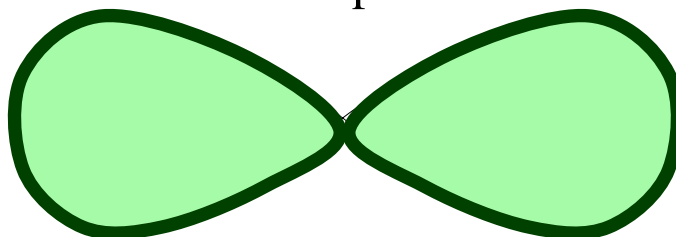
63

Stimulus:	Cells in V2
Probe location: ●	Response?
	YES
	NO
	NO
	YES
 (more contrast)	NO
	YES

von der Heydt,
Peterhans, and
Baumgartner, 1984

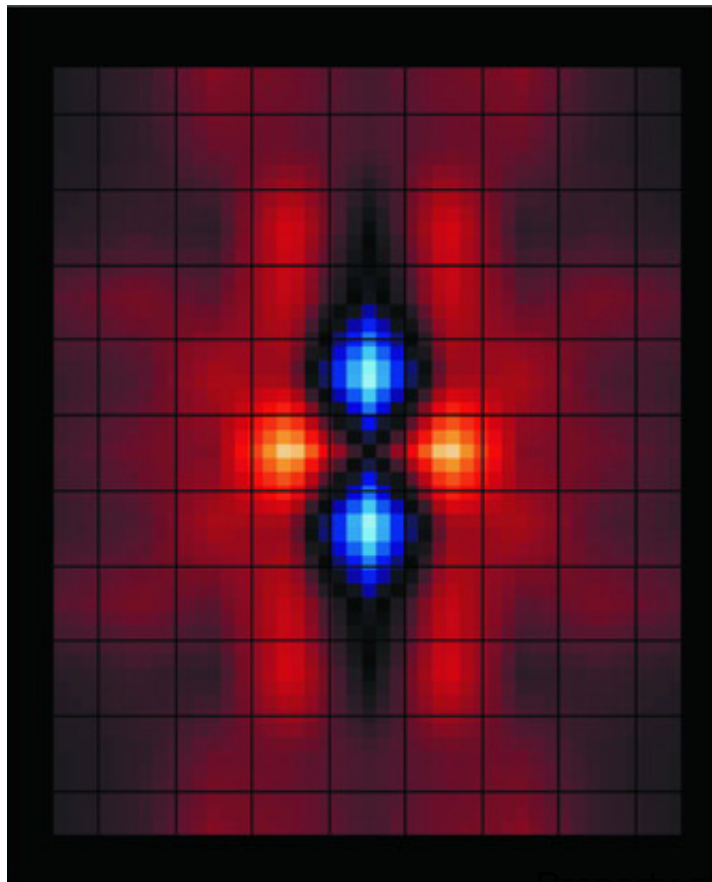
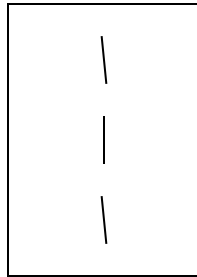
Peterhans and
von der Heydt, 1988

Evidence for receptive field:

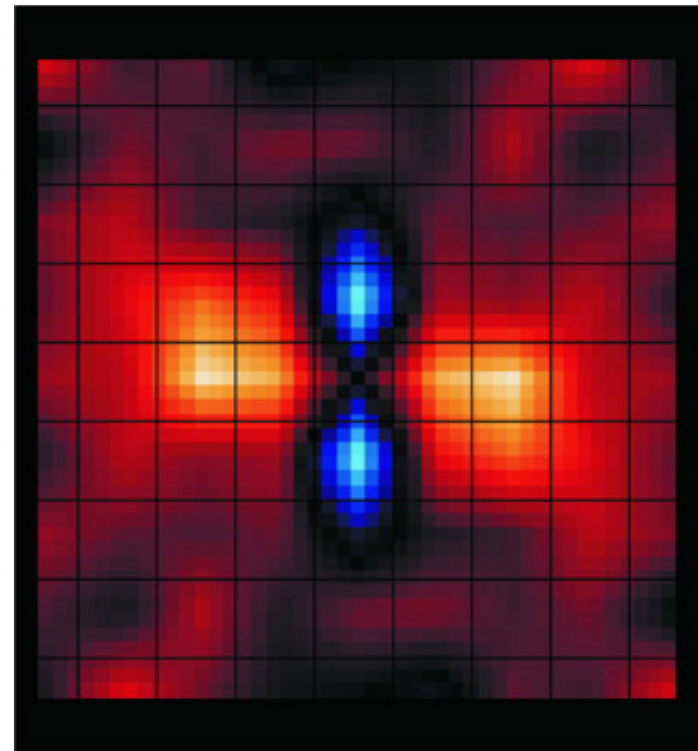


KAPADIA, ITO, GILBERT & WESTHEIMER (1995)

Psychophysics

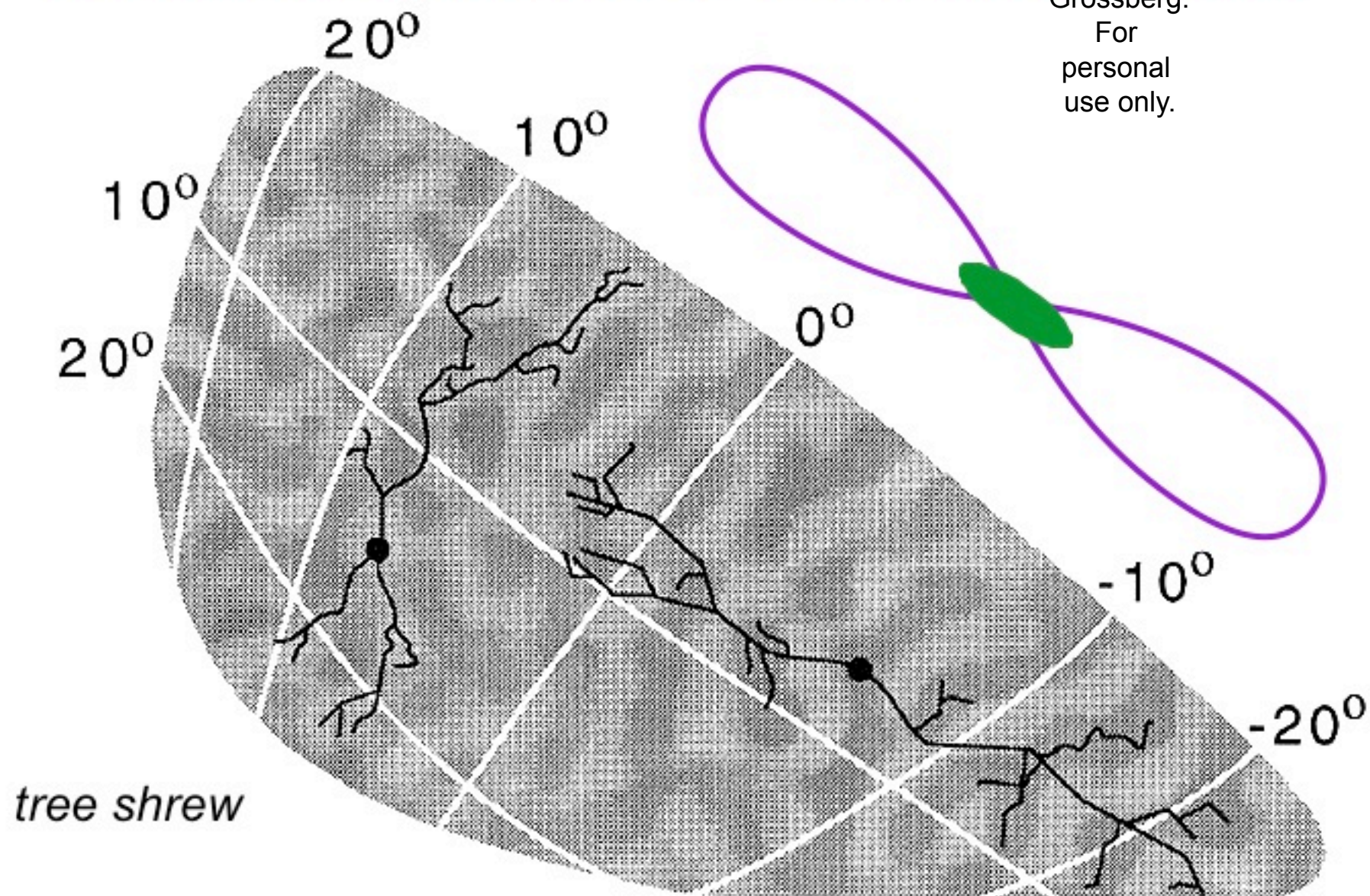


Neurophysiology
V1



ANATOMY: HORIZONTAL CONNECTIONS (V1)

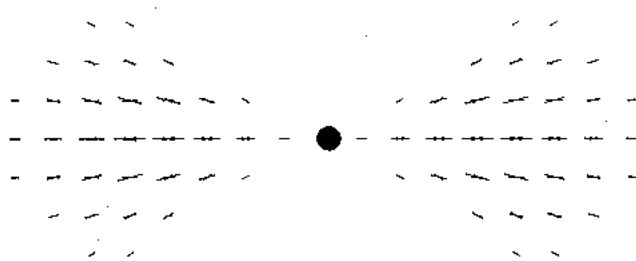
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Bosking, et al., 1997

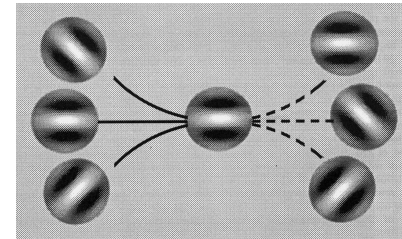
BIPOLES THROUGH THE AGES

Grossberg and Mingolla, 1985

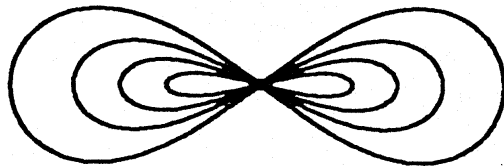


Field, Hayes, and Hess, 1993

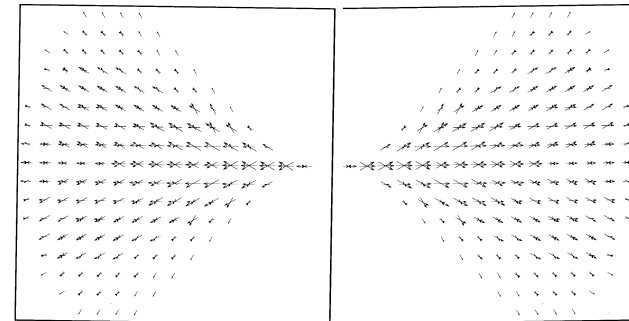
“association field”



Heitger and von der Heydt, 1993



Williams and Jacobs, 1997



Cf. “relatability” geometric constraints on which contours
get to group with which Kellman & Shipley, 1991

HOW DOES GROUPING OCCUR IN THE VISUAL CORTEX?

BREAKTHROUGHS IN BRAIN COMPUTING

Models that link detailed **BRAIN CIRCUITS** to the
ADAPTIVE BEHAVIORS that they control

Mind/Body Problem

Describe **NEW PARADIGMS** for brain computing

INDEPENDENT MODULES
Computer Metaphor

COMPLEMENTARY COMPUTING

What is the nature of brain specialization?

LAMINAR COMPUTING

Why are all neocortical circuits organized in layers?
How do laminar circuits give rise to biological intelligence?

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GROUPING STARTS IN LAYER 2/3

Long-range
horizontal excitation links
collinear, coaxial receptive fields

Gilbert & Wiesel, 1989

Bosking et al., 1997

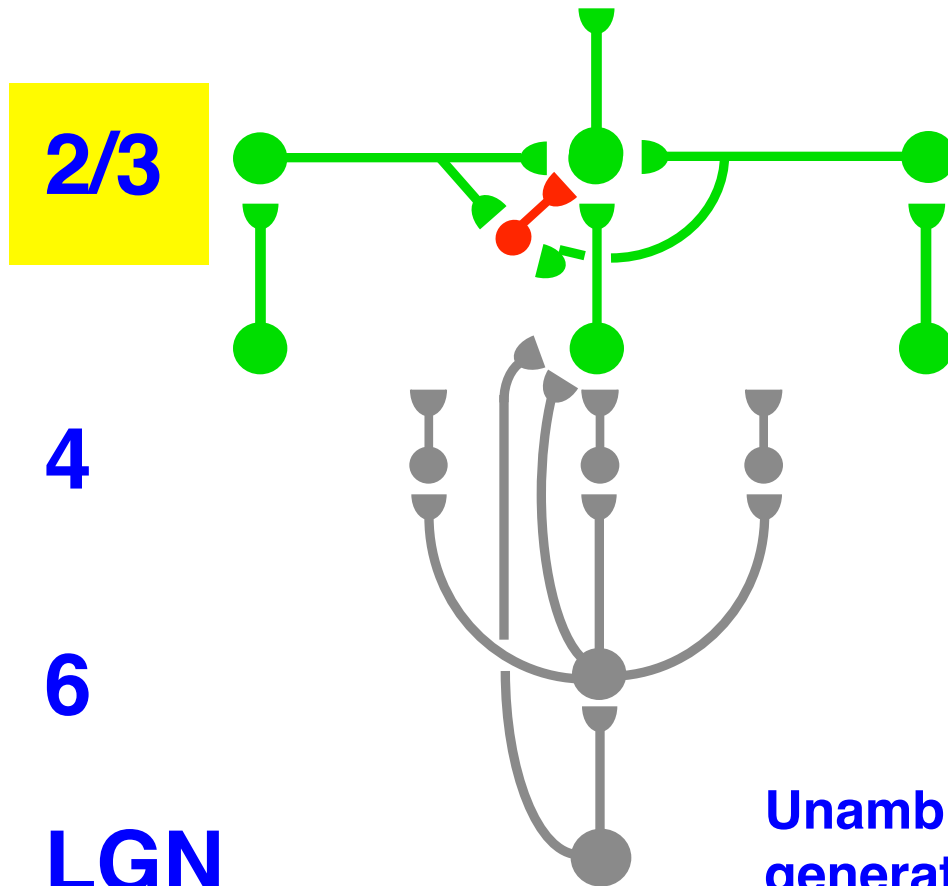
Schmidt et al, 1997

Short-range
disynaptic inhibition of target
pyramidal via pool of
interneurons

Hirsch & Gilbert, 1991

Unambiguous groupings can form and
generate feedforward outputs quickly

Thorpe et al, 1996



GROUPING STARTS IN LAYER 2/3

What happens before layer 2/3?

Long-range
horizontal excitation links
collinear, coaxial receptive fields

Gilbert & Wiesel, 1989

Bosking et al., 1997

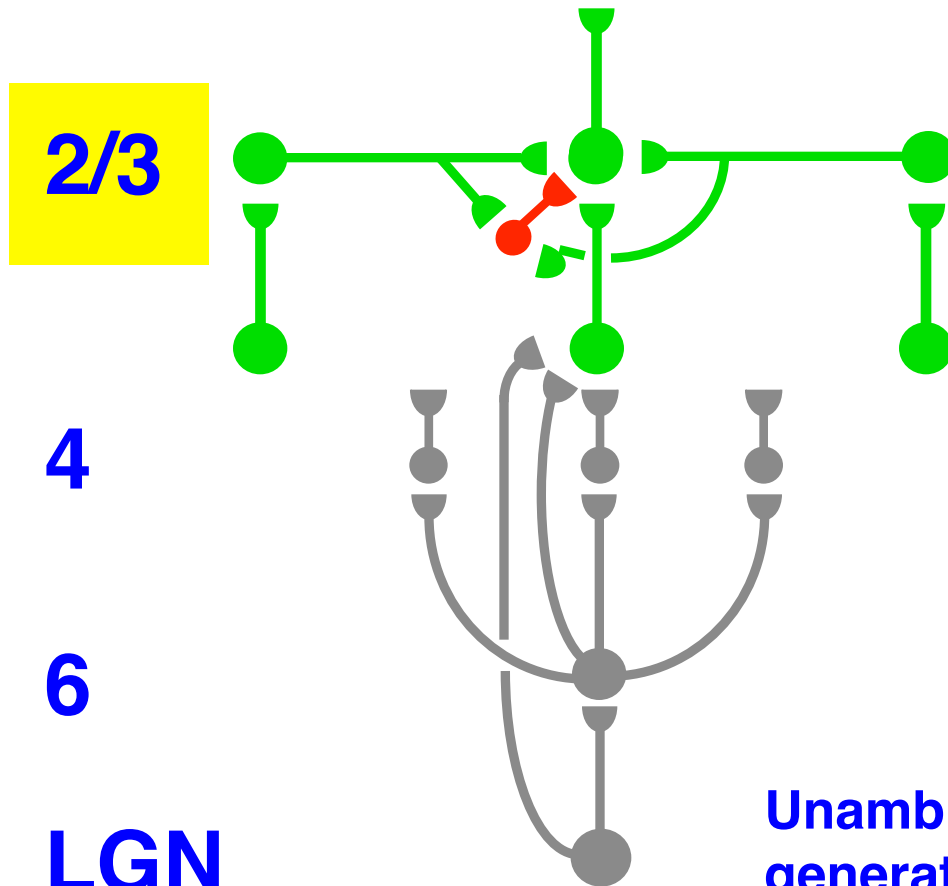
Schmidt et al, 1997

Short-range
disynaptic inhibition of target
pyramidal via pool of
interneurons

Hirsch & Gilbert, 1991

Unambiguous groupings can form and
generate feedforward outputs quickly

Thorpe et al, 1996



FROM ORIENTED FILTERING TO GROUPING AND BOUNDARY COMPLETION

Oriented Receptive Fields: SIMPLE CELLS

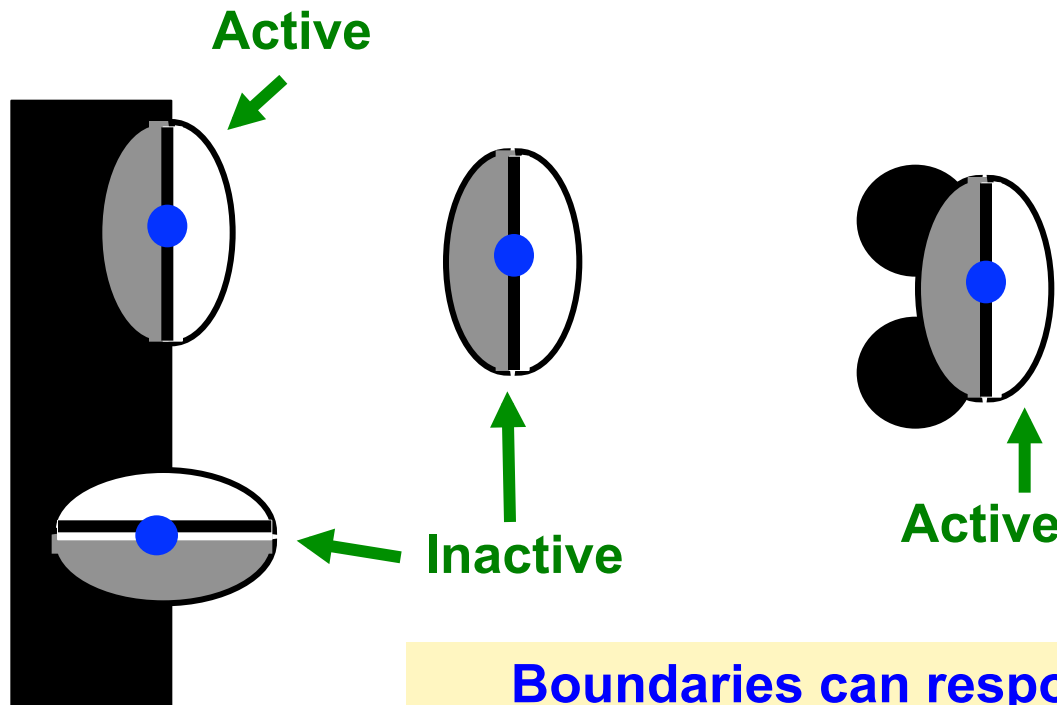
Hubel and Wiesel, 1968

Sensitive to:

orientation
amount of contrast
direction of contrast
spatial scale

**ORIENTED LOCAL
CONTRAST DETECTORS**

Not EDGE detectors!



Boundaries can respond to
edges, textures, and shading
in a form-sensitive way

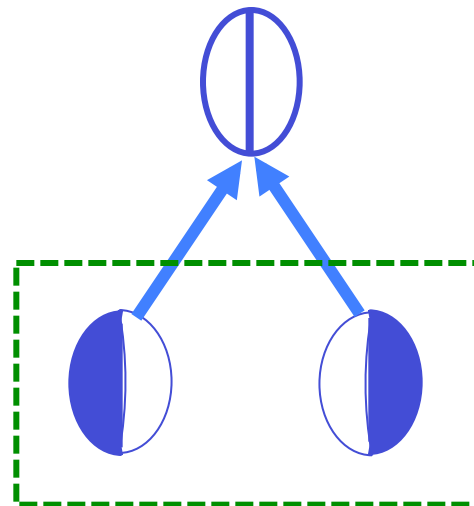
BOUNDARY WEBS!

SIMPLE CELLS ARE NOT SUFFICIENT!

ALL BOUNDARIES ARE INVISIBLE: COMPLEX CELLS

complex cells pool inputs from
opposite-polarity simple cells in V1

V1



complex cells

simple cells

Complex cells are amodal boundary detectors Grossberg (1984)

vs

“color cells in the broadest sense” Thorell, DeValois & Albrecht (1984)

COMPLEX CELLS ARE NOT SUFFICIENT!

Property of
Stephen
Grossberg.

For
personal
use only.

They cannot detect places in an image or painting with
sudden curvature changes

Corners

Line ends



BRAINS DO NOT OBEY CLASSICAL GEOMETRY

A line is NOT A COLINEAR SERIES OF POINTS

**A line is an
EMERGENT PROPERTY
of
MULTIPLE PROCESSING STAGES
that realize a**

HIERARCHICAL RESOLUTION OF UNCERTAINTY

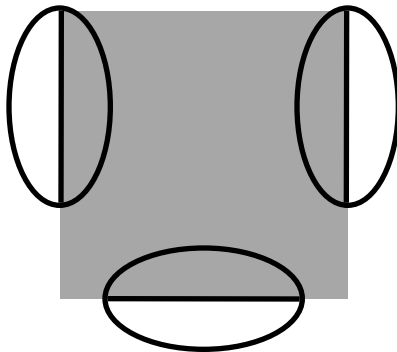
HIERARCHICAL RESOLUTION OF UNCERTAINTY

For a given receptive field size:

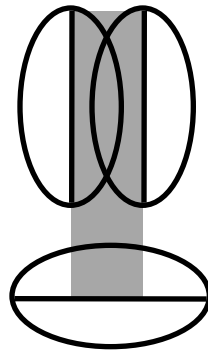


Different responses occur at bar ends and line ends:

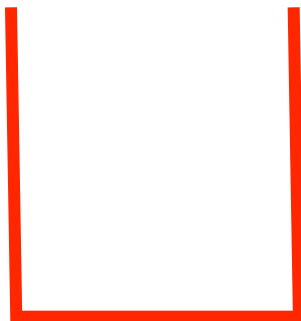
bar end



line end



For a *thin* line
no detector
perpendicular to line end
can respond enough to
close the boundary there



Network activity

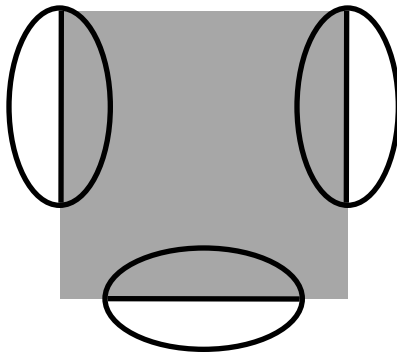
HIERARCHICAL RESOLUTION OF UNCERTAINTY

For a given receptive field size:

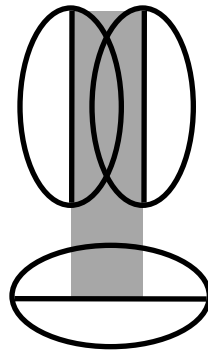


Different responses occur at bar ends and line ends:

bar end



line end



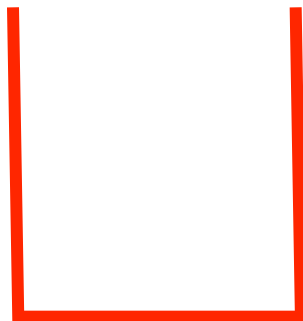
For a *thin* line

no detector

perpendicular to line end

can respond enough to

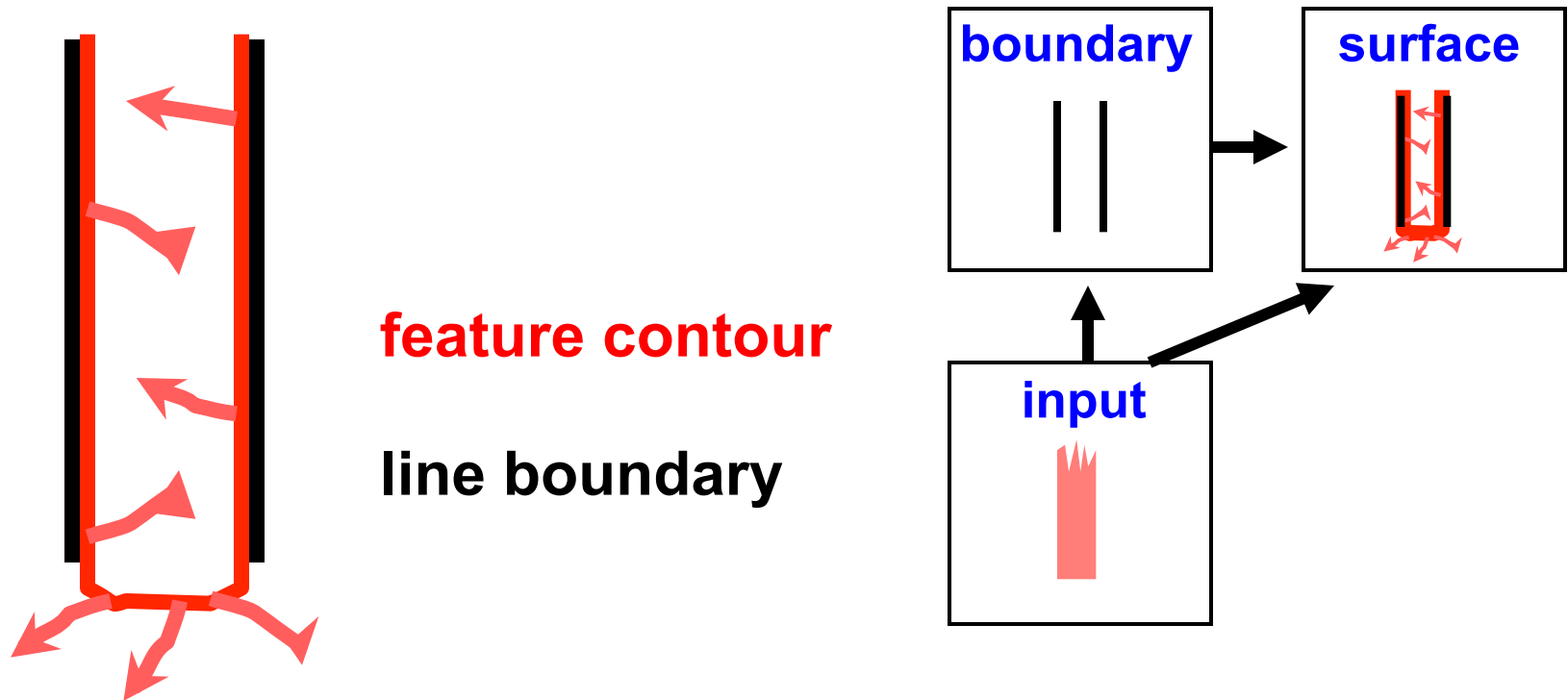
close the boundary there



Network activity

WHO CARES?!

IT WOULD CAUSE A PERCEPTUAL DISASTER WHEN SURFACE FILLING-IN OCCURS



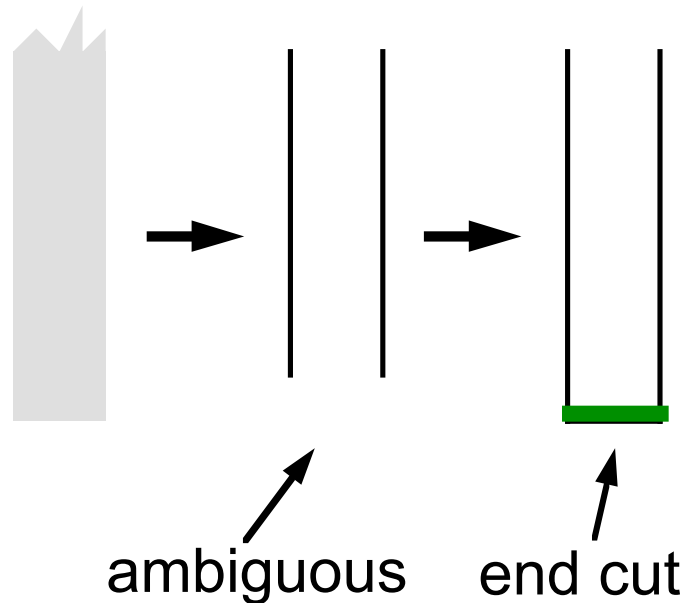
Color would flow from EVERY line end!
...as it does during neon color spreading

HIERARCHICAL RESOLUTION OF UNCERTAINTY

END CUTS

The boundary system must **CREATE** a line end at next processing stage:

EVERY LINE END IS ILLUSORY!

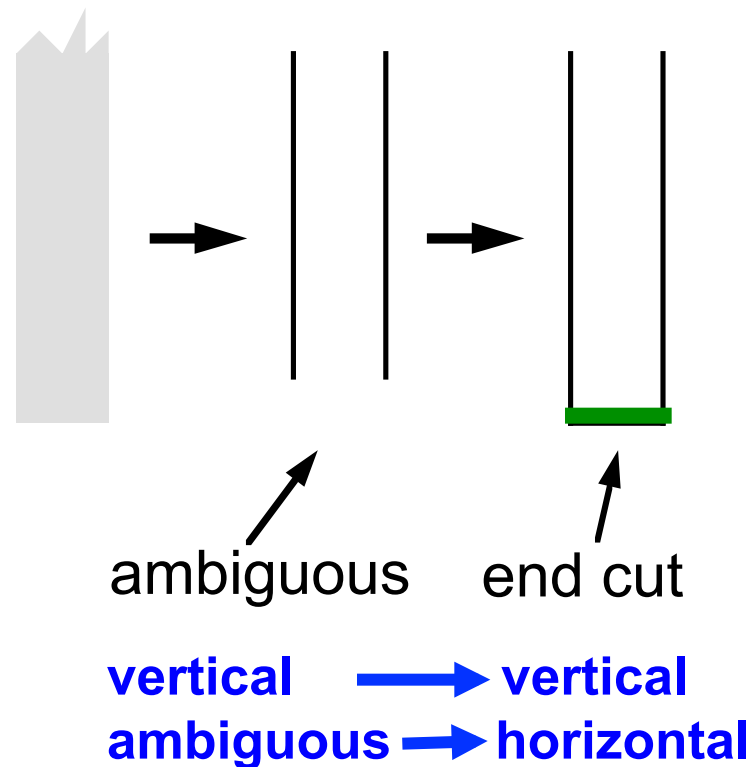


HIERARCHICAL RESOLUTION OF UNCERTAINTY

END CUTS

The boundary system must **CREATE** a line end at next processing stage:

EVERY LINE END IS ILLUSORY!



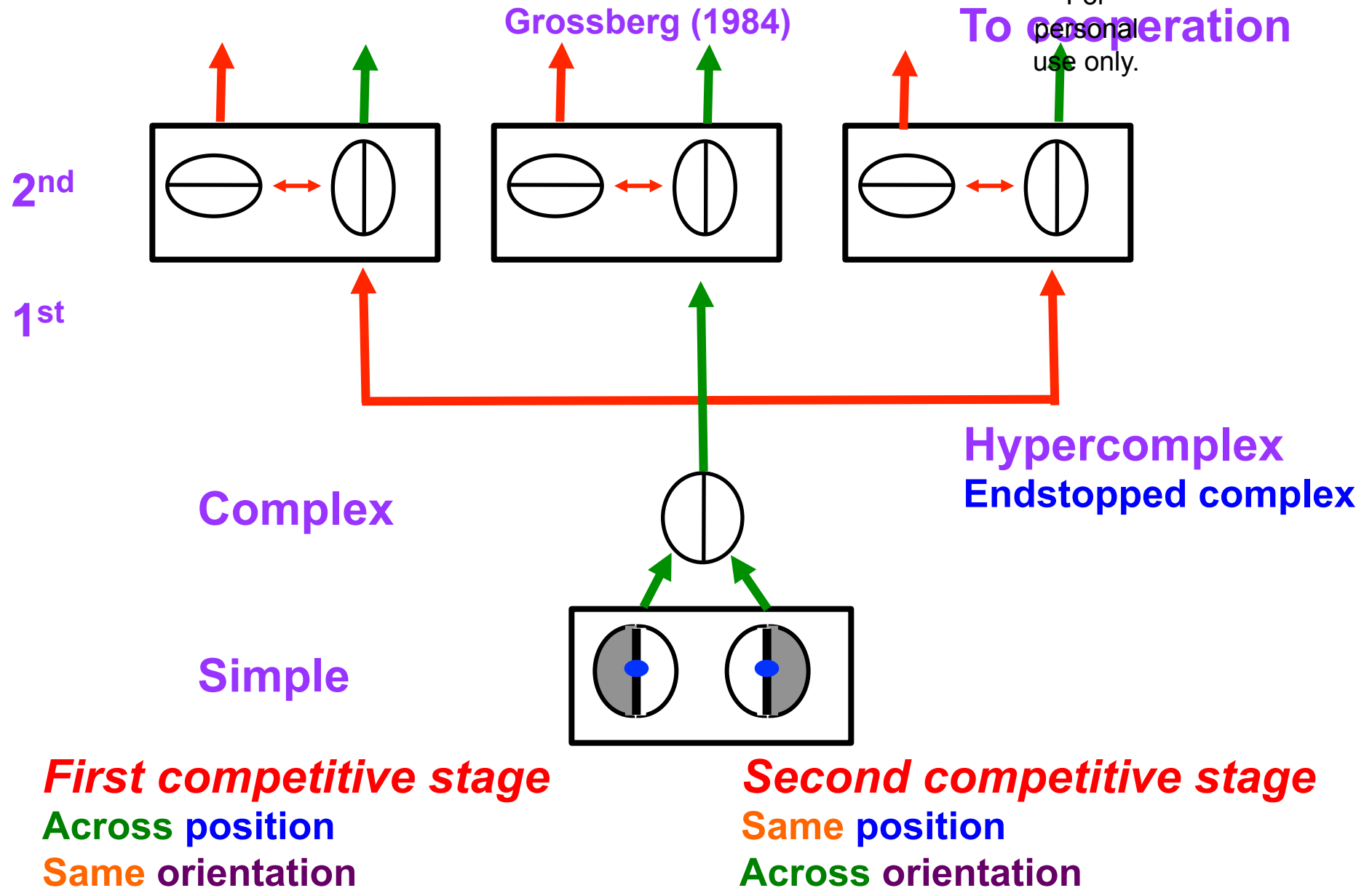
A Pattern-to-Pattern Map
Not a Pixel-to-Pixel Map

HOW ARE END CUTS CREATED?

Two stages of short-range competition

Property of
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Grossberg.
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81



HOW END CUTS ARE CREATED

FIRST COMPETITIVE STAGE

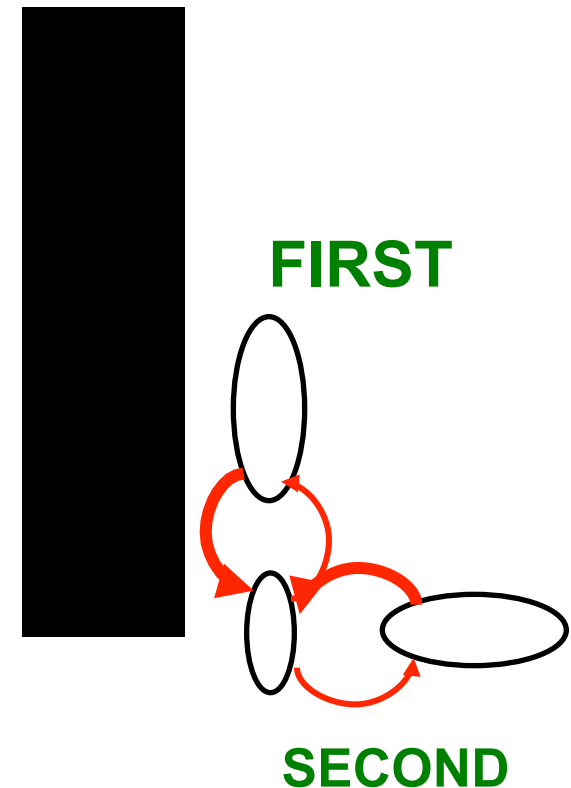
Vertically oriented complex cells near line end inhibit vertically oriented hypercomplex cells just beyond the line end

These hypercomplex cells were not previously active, so at the...

SECOND COMPETITIVE STAGE

Inhibited vertically oriented hypercomplex cells near the line end **disinhibit** horizontally oriented hypercomplex cells at the same positions, thereby creating an **end cut**

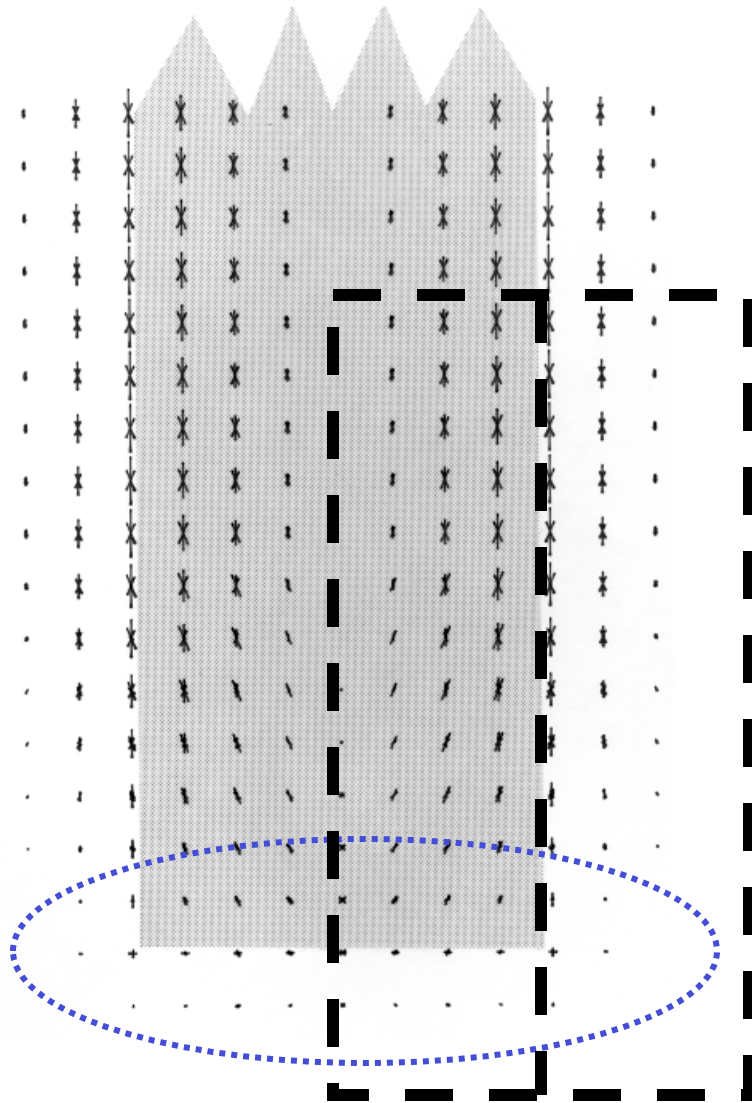
Cells in second competitive stage are **TONICALLY** active but are held in check by balanced inhibition until an input occurs



END CUT SIMULATION

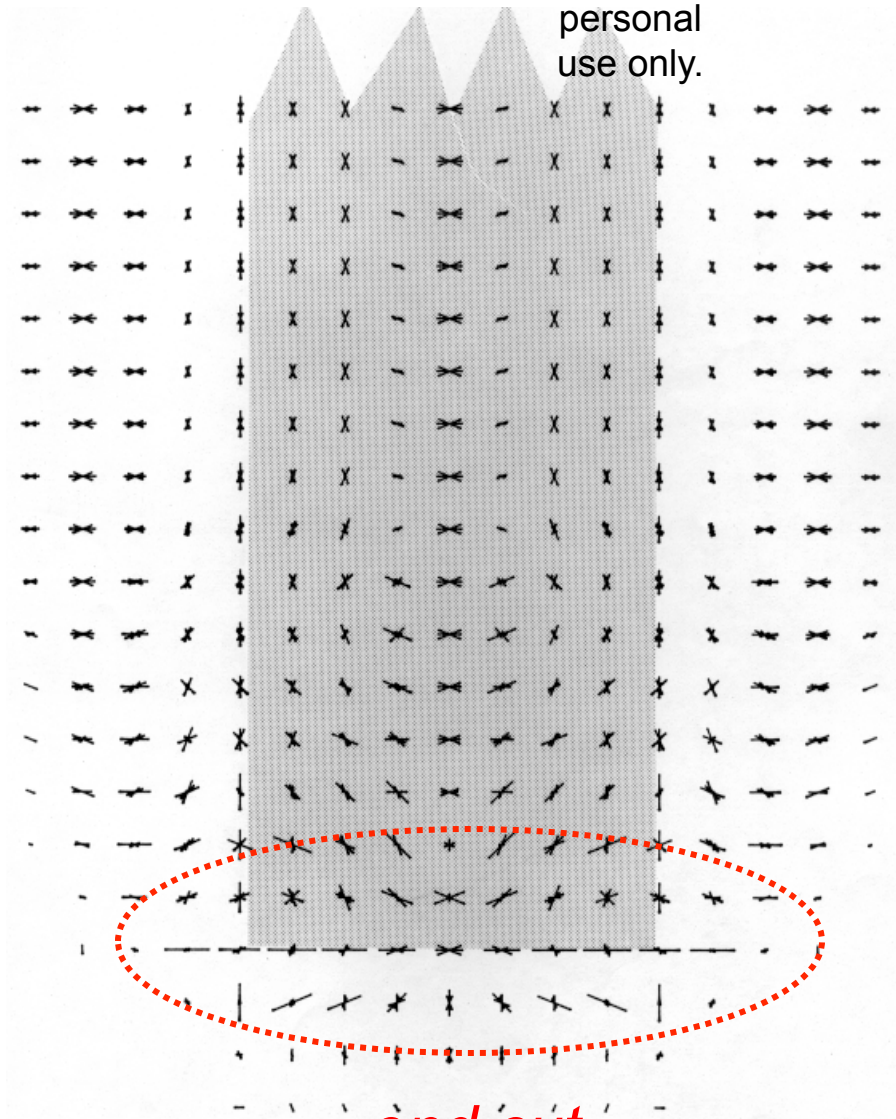
Grossberg and Mingolla (1985)

Property of
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Grossberg.
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personal
use only.



no line end

filter size



end cut

OUR BRAINS TRY TO MAKE THEIR OWN SERIFS!

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Stephen
Grossberg.

For
personal
use only.

E.g., Times, Times New Roman

AaBbCc

AaBbCc

“Serifed fonts are widely used for body text because they are considered easier to read than sans-serif fonts in print.”

Serifs, Wikipedia

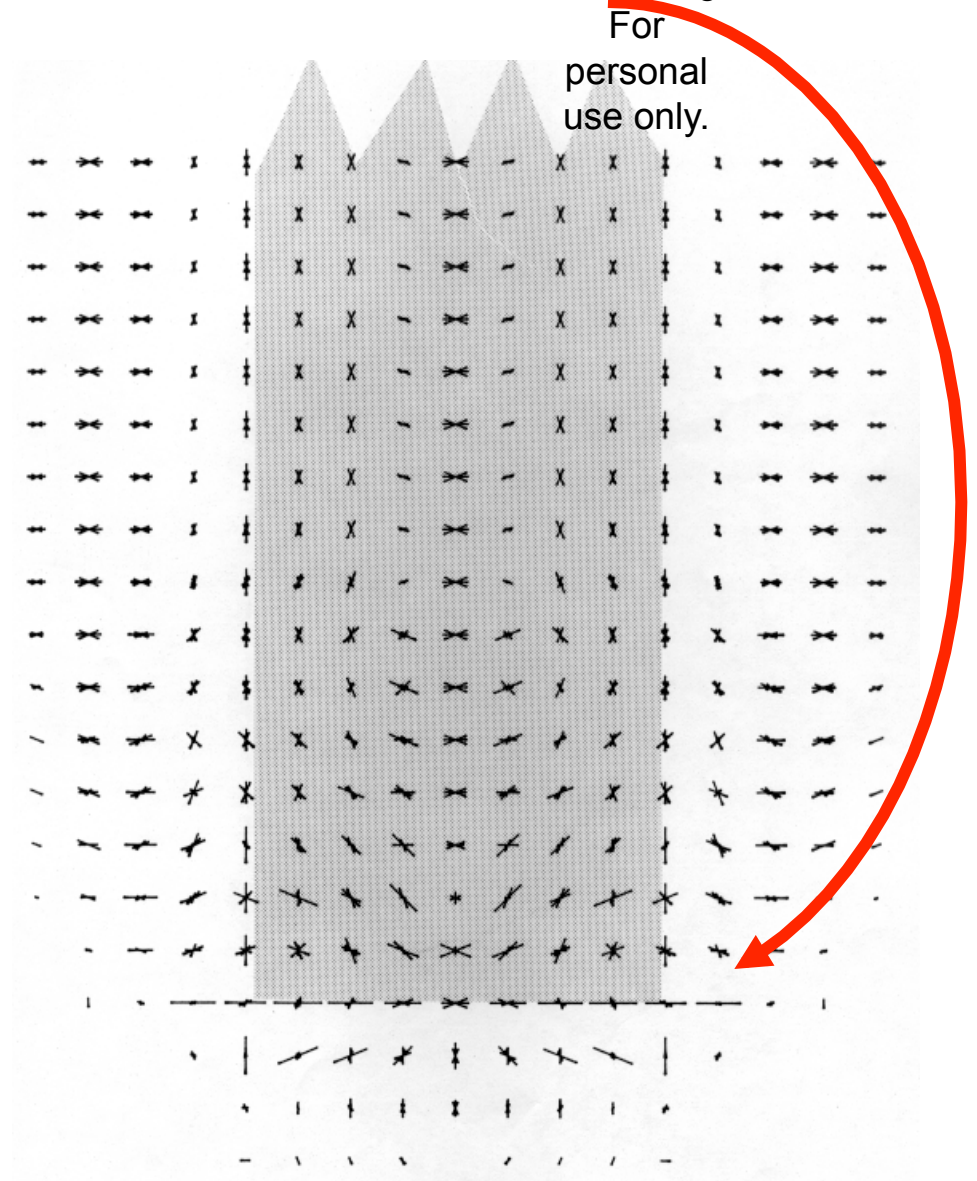
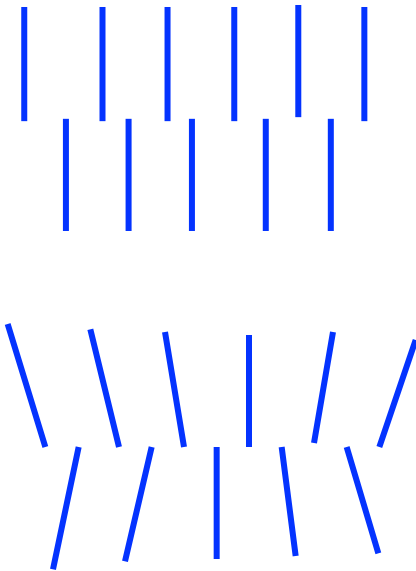
POSITIONAL HYPERACUITY, ORIENTATIONAL FUZZINESS

Property of
Stephen
Grossberg.

85

Why is orientational fuzziness useful?

It lets groupings occur at orientations that are not too different from vertical:

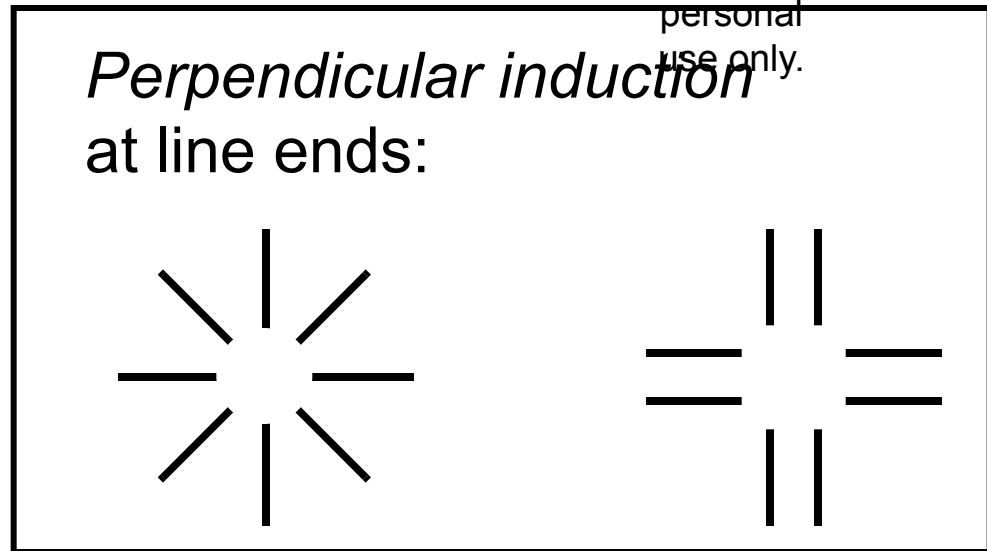


GROUPING AT PREFERRED AND UNPREFERRED ORIENTATIONS

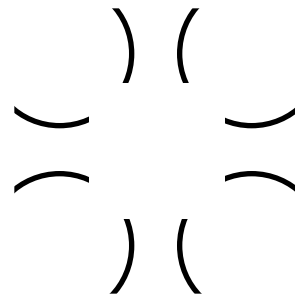
Property of
Stephen
Grossberg.
For

86

Locally preferred
and
globally preferred



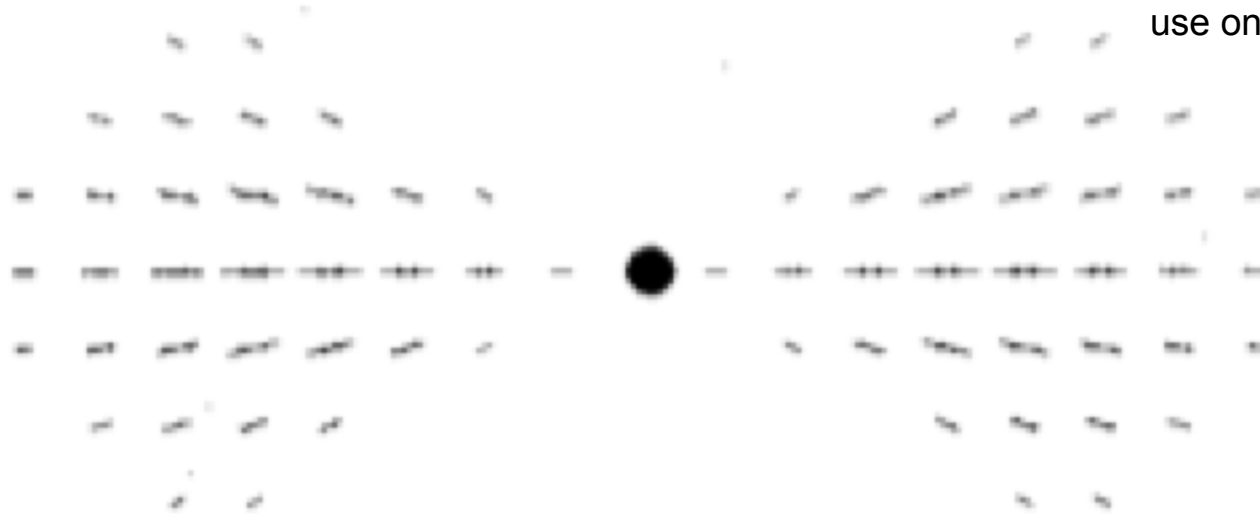
Locally unpreferred
and
globally preferred



Kennedy, 1979

Global grouping can overcome local preferences

BIPOLE CELLS HAVE FUZZY RECEPTIVE FIELDS



Why are not all the groupings that they form FUZZY?

If they were, acuity would go down a lot!

FROM FUZZY TO SHARP

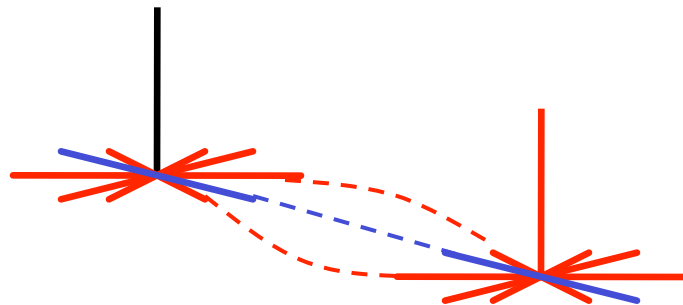
Why do we not always perceive fuzzy illusory contours?

Another **hierarchical resolution of uncertainty**:
Need fuzziness to initiate grouping
But this risks loss of acuity

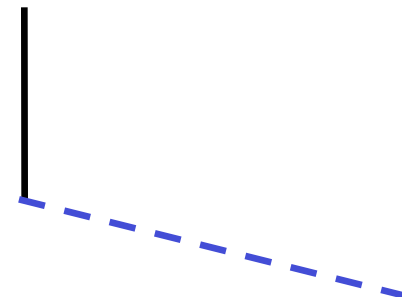
BOUNDARY GROUPING PROCESS MAKES DECISION

CHOOSE: the contextually best orientation – **cooperation**

SUPPRESS: other local orientations – **competition**



before choice
(transient)

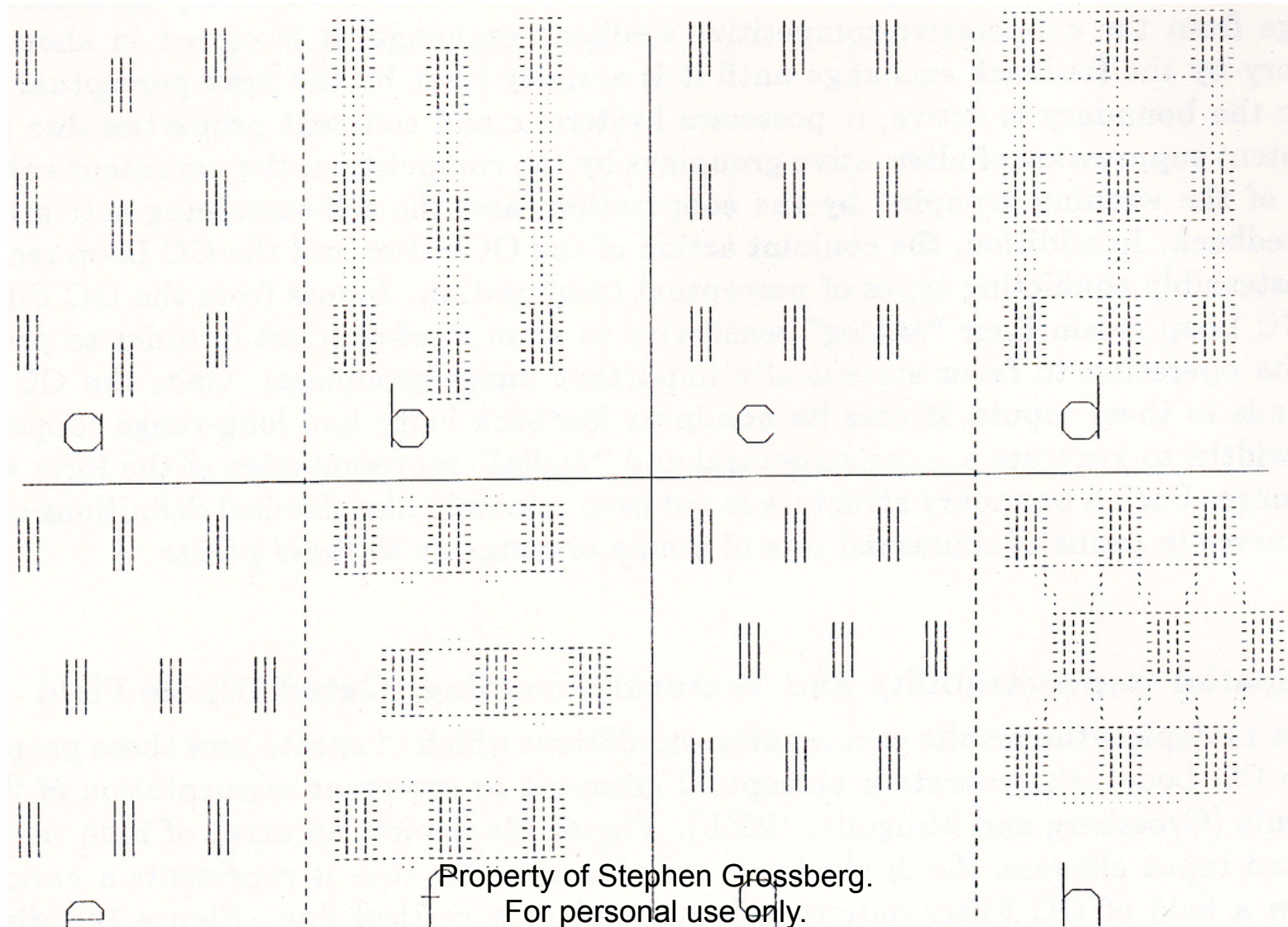


after choice
("equilibrium")

BIPOLE CELLS DO THIS
when they are part of a larger network

COMPUTER SIMULATIONS OF BOUNDARY COMPLETION

Grossberg and Mingolla (1985)



Property of Stephen Grossberg.
For personal use only.

MULTIPLE ORIENTATIONS CAN HEREBY INDUCE SHARP BOUNDARY COMPLETION OF AN OBJECT

diagonal



AAAAAA
AAAAAA
AA
AAAAAA
AAAAAA
AA
AAAAAA
AAAAAA

perpendicular



```

HHHHHHHH
H
H
HHHHHHHH
                H
                H
HHHHHHHH

```

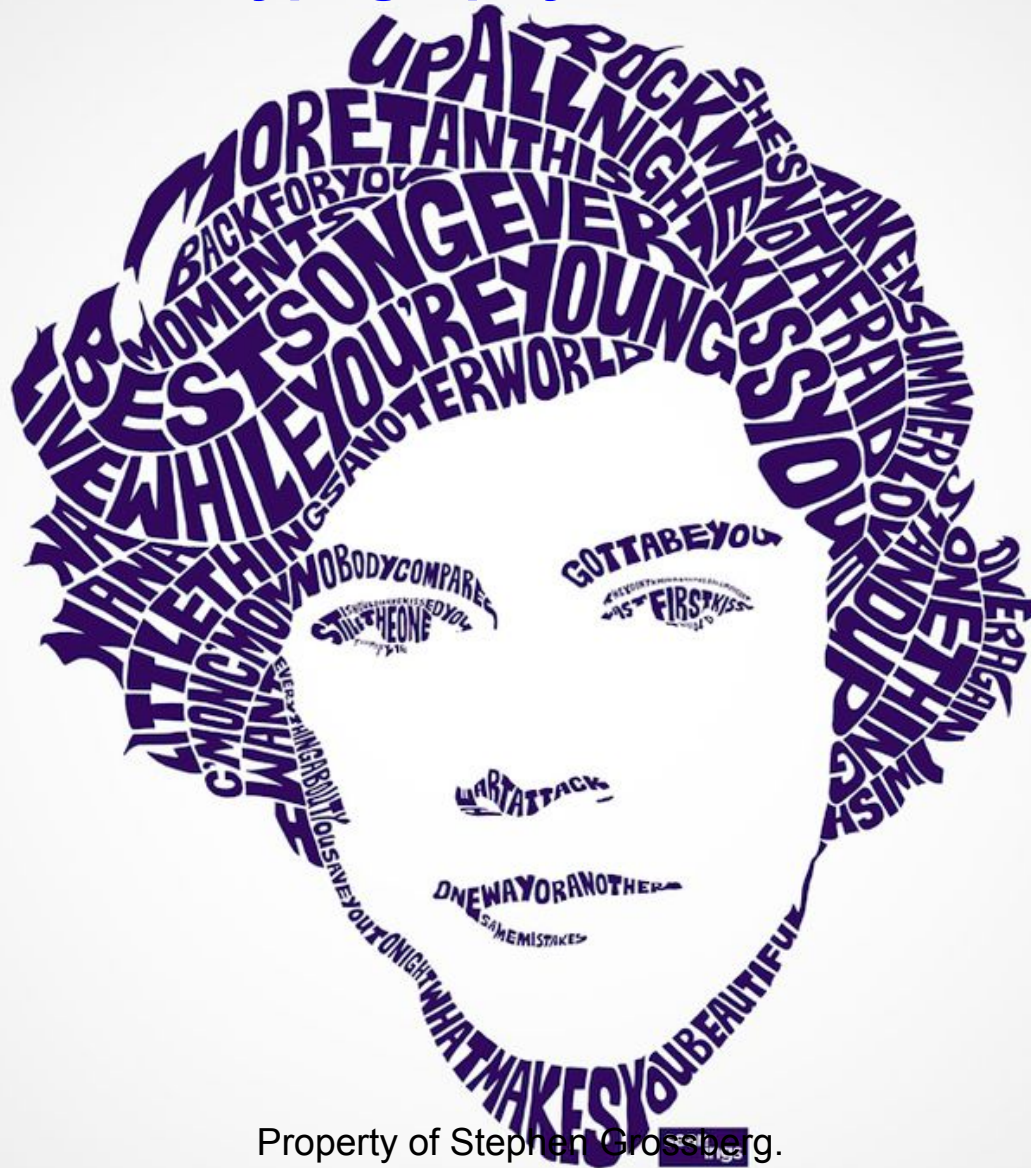
parallel



Free

SEAN WILLIAMS

Typography Portrait



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For personal use only.

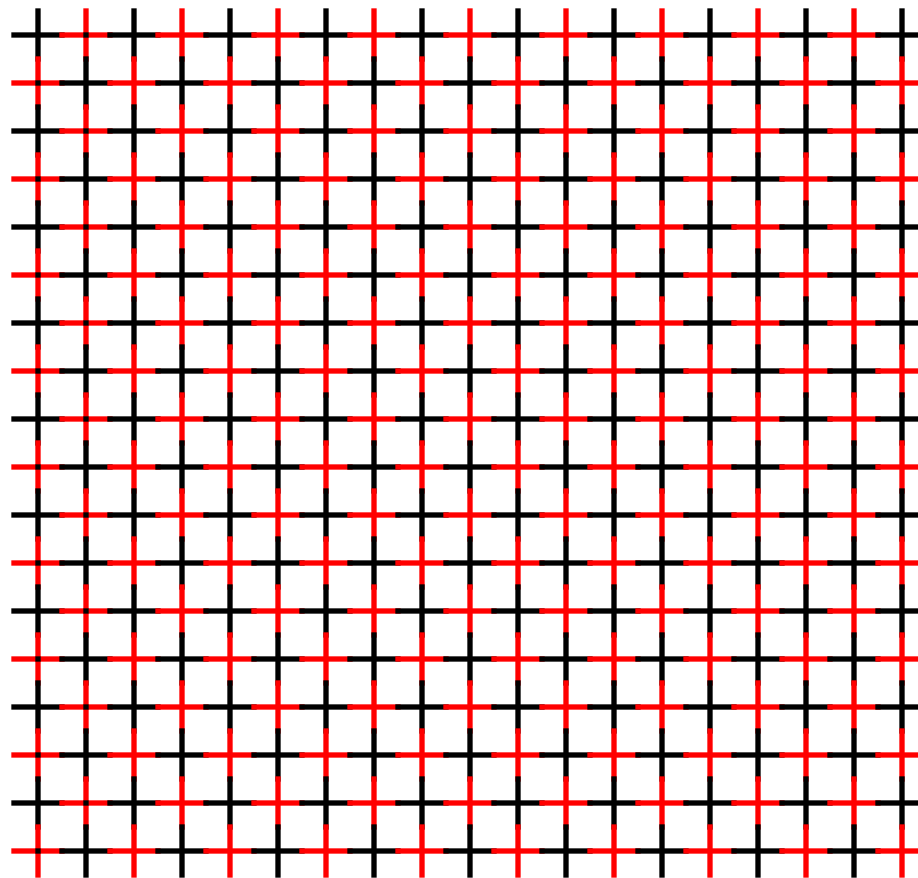
THIS CIRCUIT CAN ALSO EXPLAIN...

NEON COLOR SPREADING

Redies and Spillmann (1981)

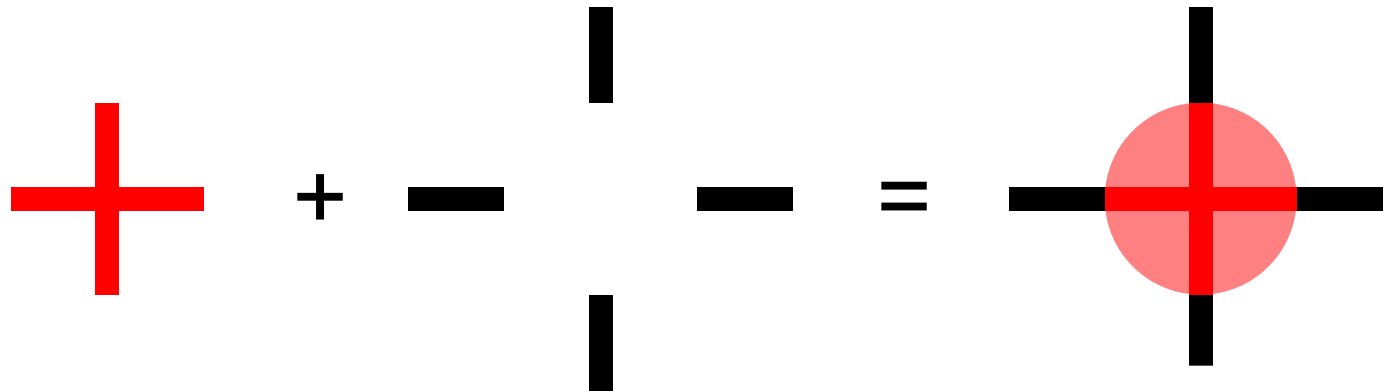
Property of
Stephen
Grossberg.
For
personal
use only.

Visible evidence for how groupings form and contain filling-in



REAL BOUNDARIES BREAK ILLUSORY BOUNDARIES CONTAIN FILLING-IN

A less contrastive red cross inside an Ehrenstein figure



produces color spreading

“Real” contours of small cross cannot contain red color

“Illusory” contours of Ehrenstein figure *do!*

HOW ARE END CUTS CREATED?

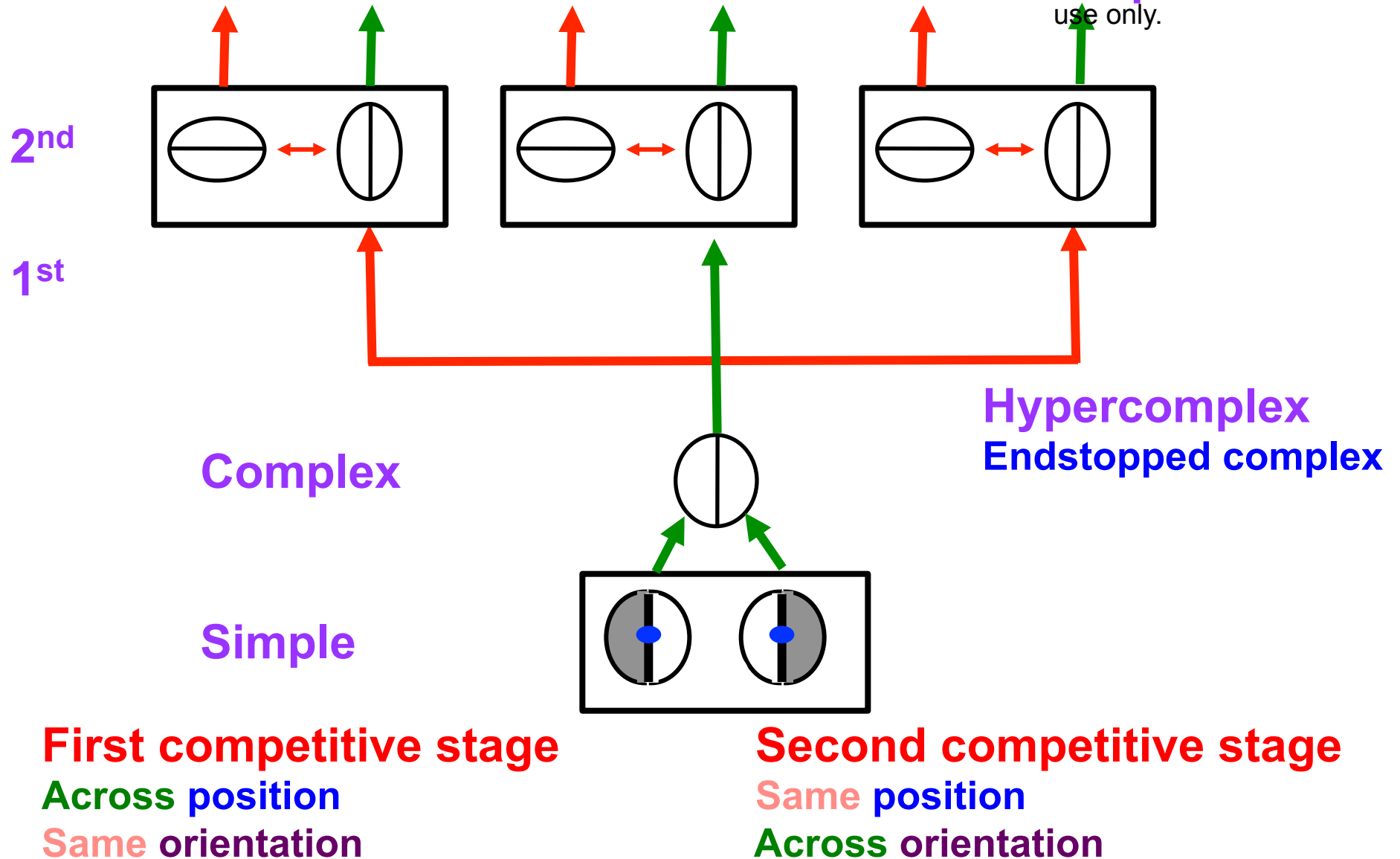
Two states of short-range competition

Property of
Stephen
Grossberg.
For
personal
use only.

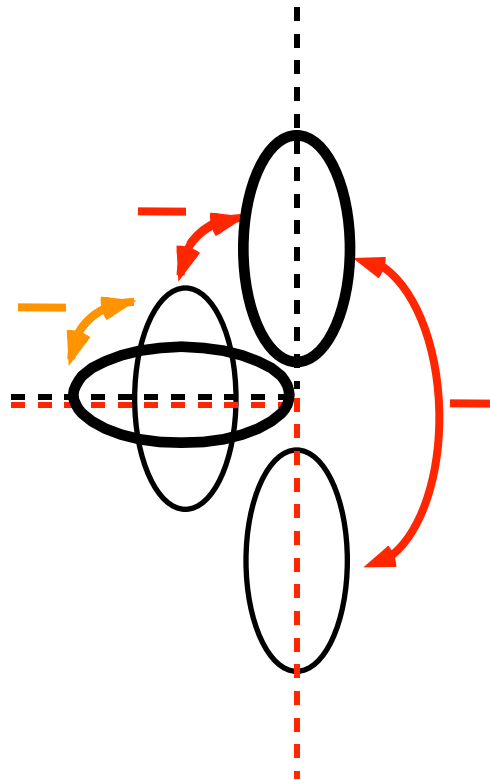
96

Grossberg, 1984

To cooperation



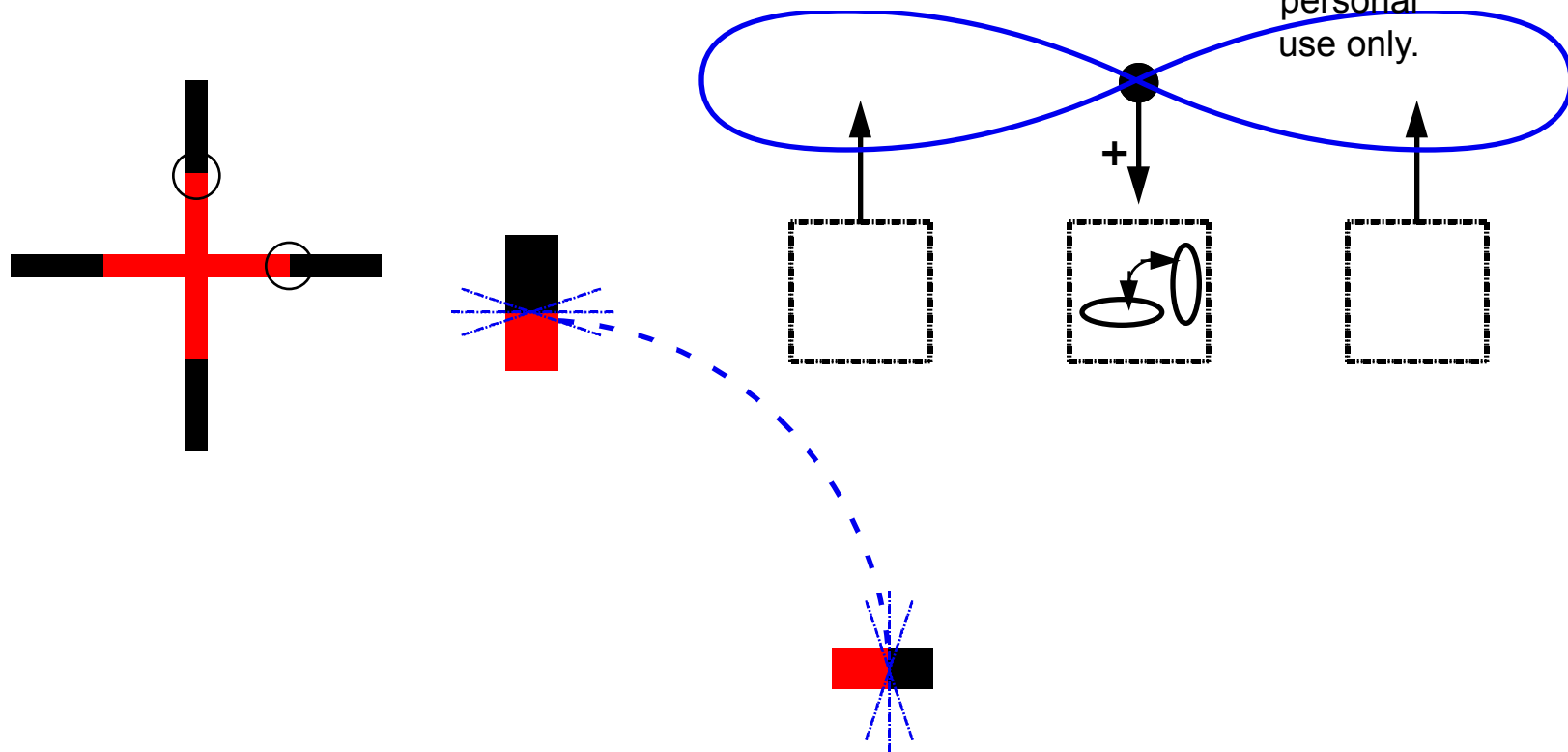
END CUT: COMPLETING BOUNDARY AT LINE END



FIRST COMPETITIVE STAGE
within orientation
across position
end gaps

SECOND COMPETITIVE STAGE
across orientation,
within position
end cuts

BIPOLE CELLS: BOUNDARY COMPLETION



long-range cooperation and short-range competition
complete winning boundary groupings
and suppress weaker boundaries

WHAT IS THIS TALK ABOUT?

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JO BAER

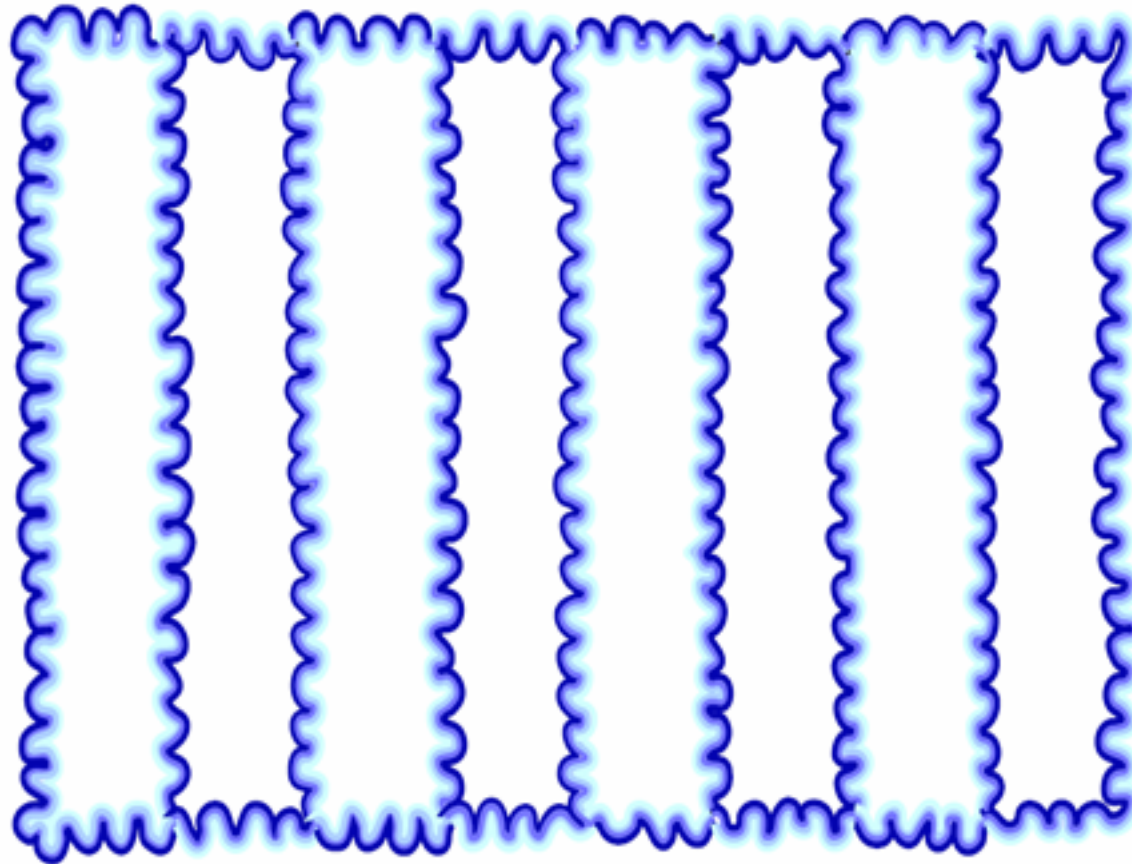
Primary Light Group: Red, Green, and Blue
1964-1965 (moma.org)



BAINGIO PINNA

Watercolor illusion, 1987

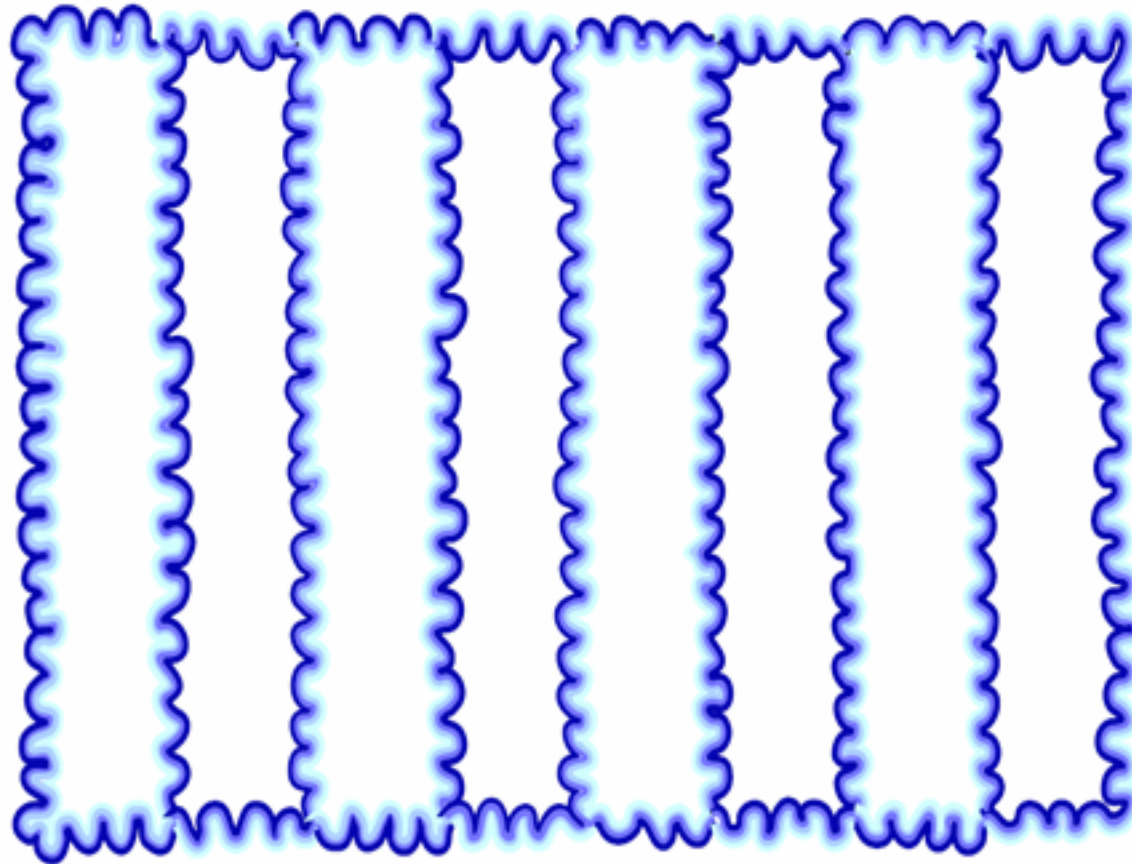
Explain using **first competitive stage**:
Spatial competition within orientation



BAINGIO PINNA

Watercolor illusion, 1987

Explain using **first competitive stage**:
Spatial competition within orientation



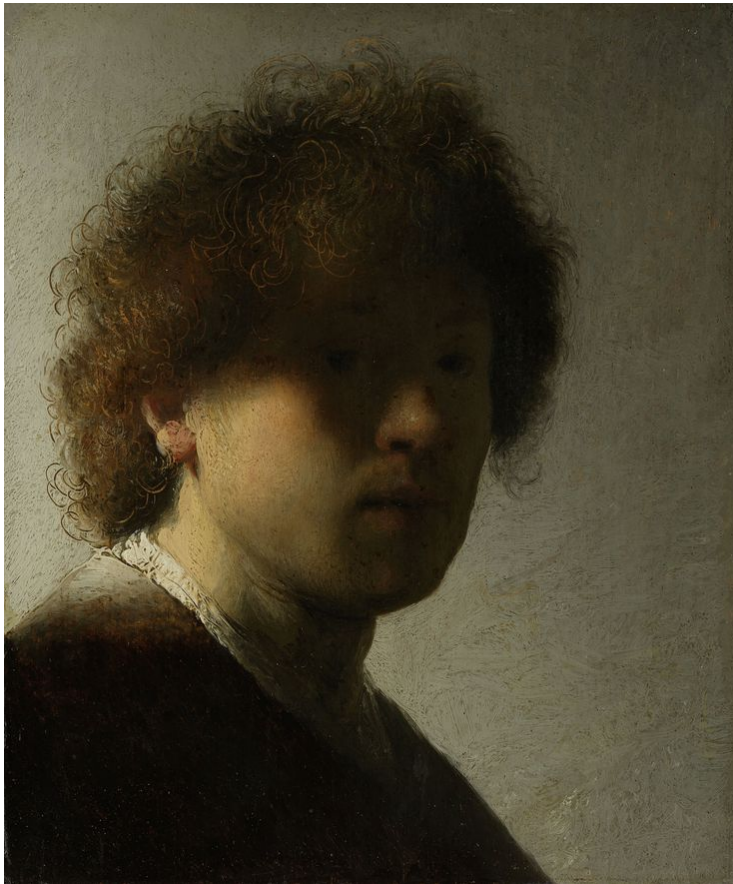
But why do the colored regions seem to bulge in depth?

Property of Stephen Grossberg.
For personal use only.

Same question for...

CHIAROSCURO

Rembrandt self-portrait



TROMPE L'OEIL

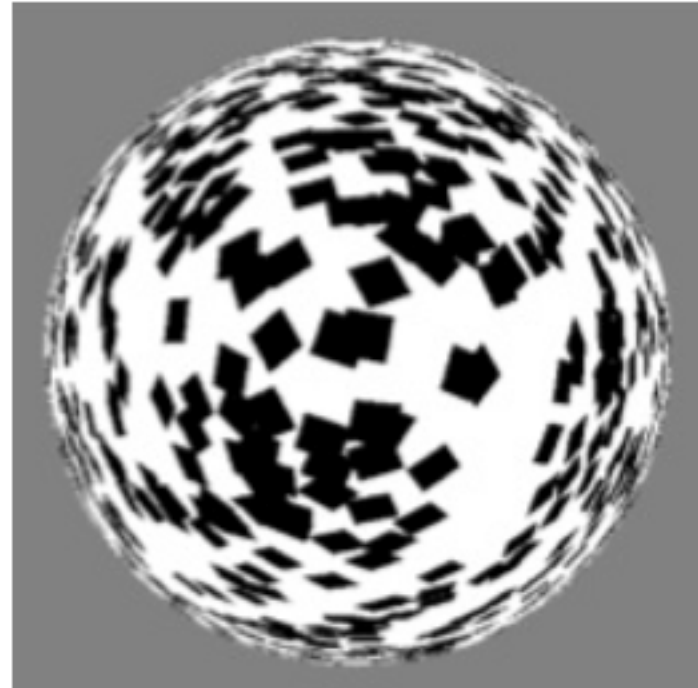
Graham Rust



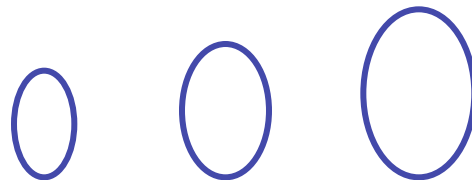
HOW CAN 2D PICTURES LOOK 3D?!

Property of Stephen Grossberg.
For personal use only.

Same question for many SHADED AND TEXTURED OBJECTS



We need receptive fields with MULTIPLE SIZES



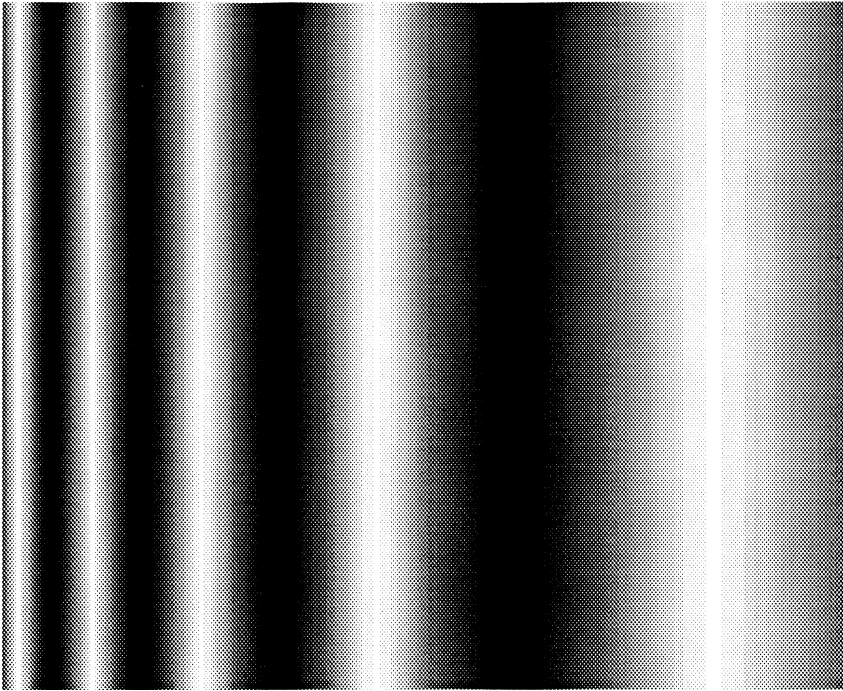
to create boundary webs that represent

MULTIPLE DEPTHS

Property of Stephen Grossberg
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WHY?

As an object approaches, it gets bigger on the retina



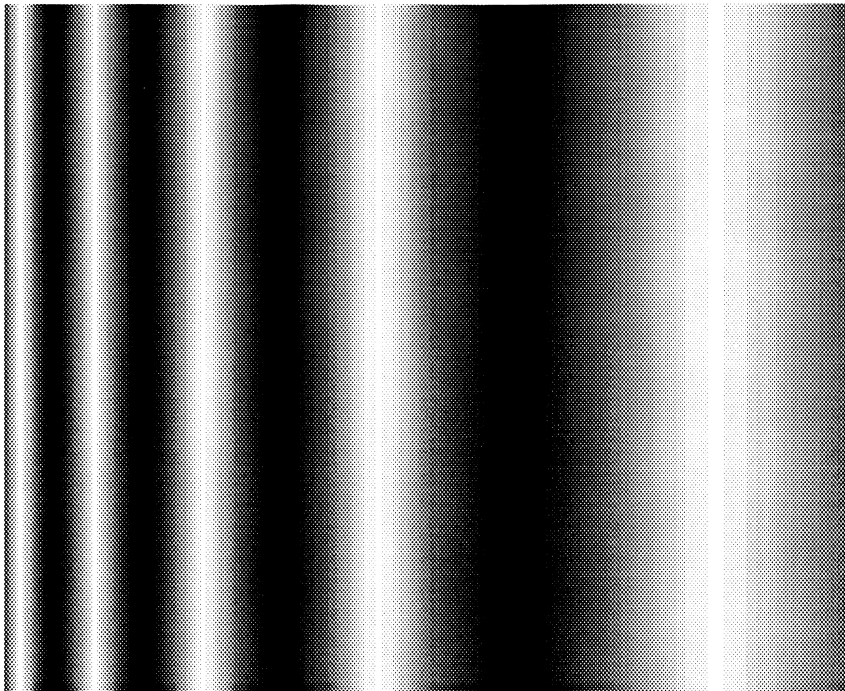
FAR

NEAR

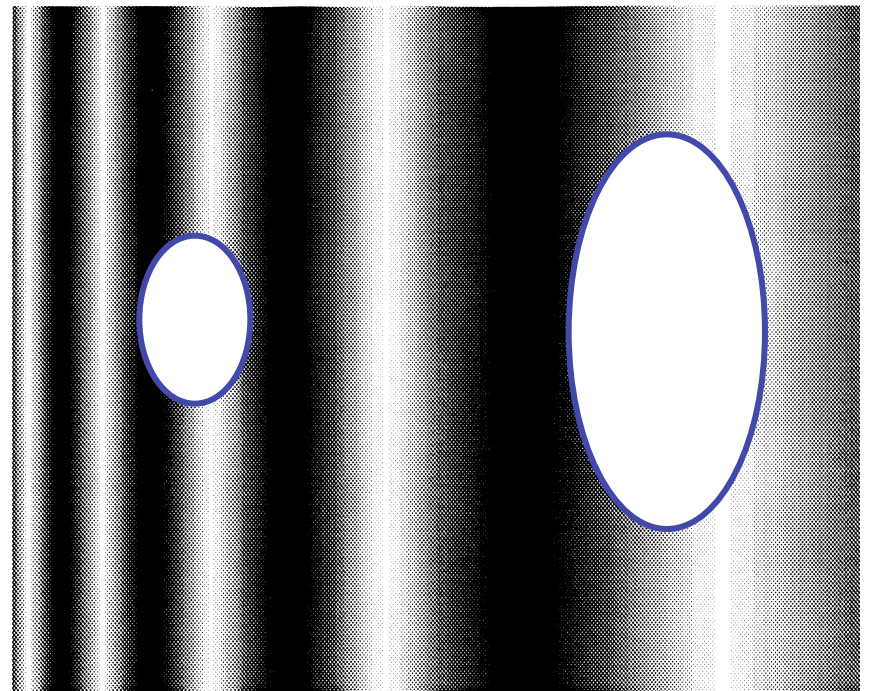
WHY?

As an object approaches, it gets bigger on the retina

That is why BIGGER scales learn to code NEARER depths

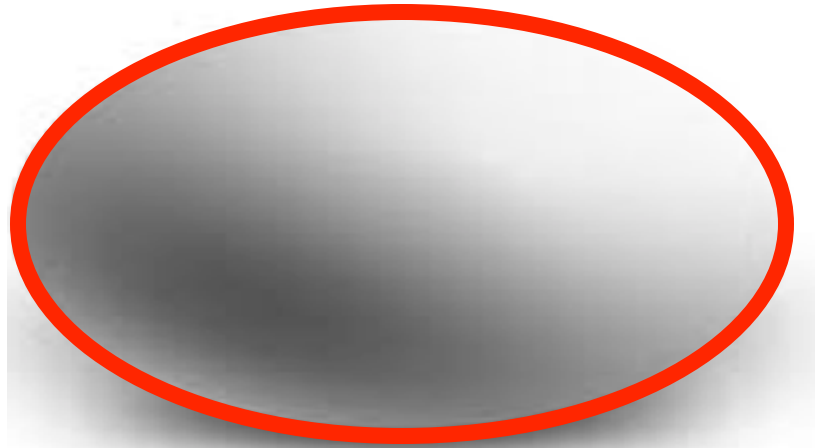


FAR



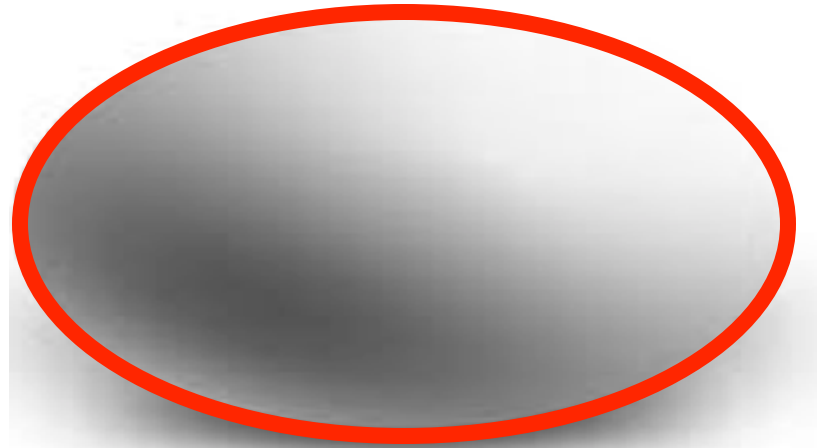
NEAR

3D VISION uses
Multiple-scale, depth-selective boundary webs

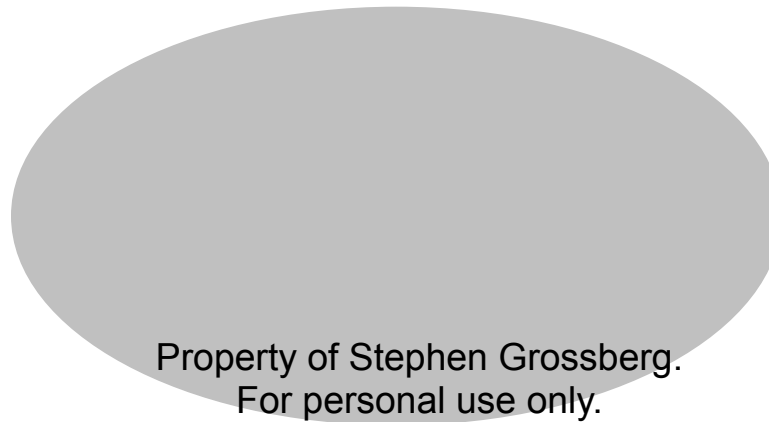


If boundaries were just edge detectors,
there would be just a bounding edge of the ellipse

3D VISION uses
Multiple-scale, depth-selective boundary webs



If boundaries were just edge detectors,
there would be just a bounding edge of the ellipse
After filling-in of the shading, it would look like this:



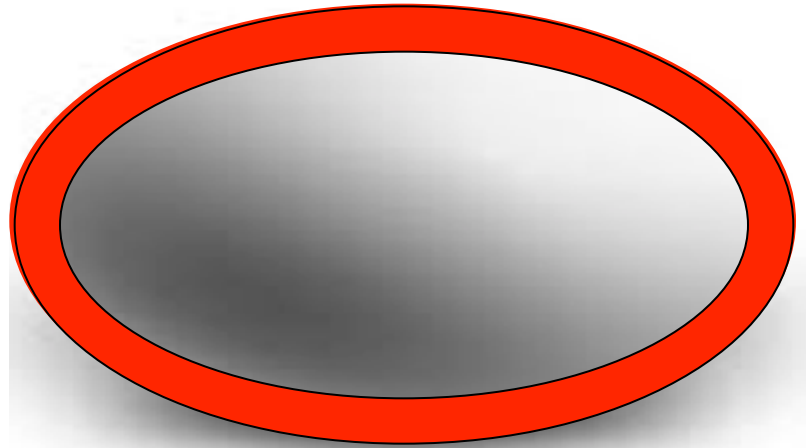
Looks flat!

3D VISION uses
Multiple-scale, depth-selective boundary webs



Instead, **different size detectors** generate dense boundary webs at different positions and **depths** along the shading gradient

3D VISION uses
Multiple-scale, depth-selective boundary webs



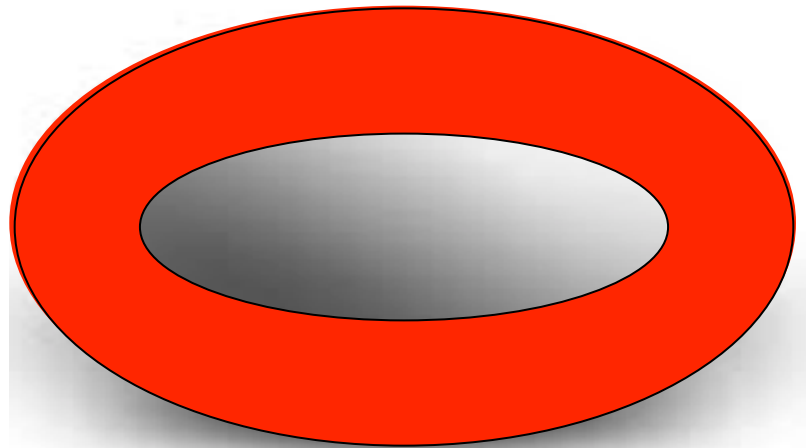
Small size

Far

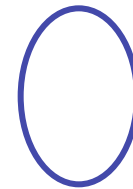


Instead, different size detectors generate dense boundary webs at different positions and depths along the shading gradient

3D VISION and Multiple-scale, depth-selective boundary webs

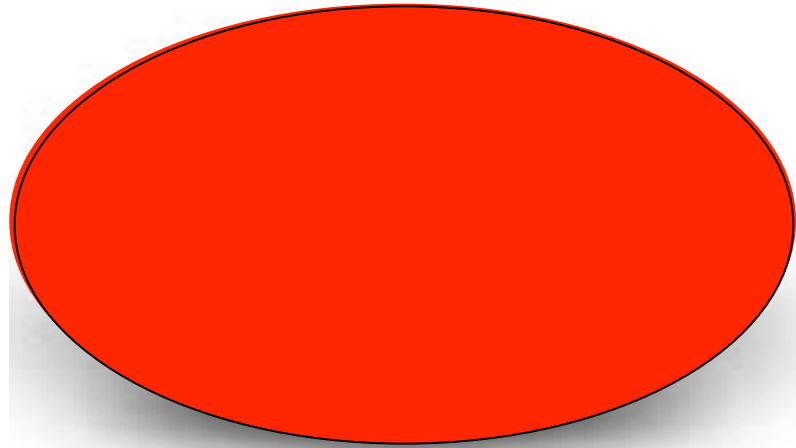


Larger size
Nearer

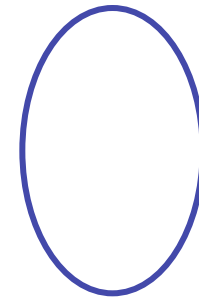


Instead, different size detectors generate dense boundary webs at different positions and depths along the shading gradient

3D VISION uses
Multiple-scale, depth-selective boundary webs



Largest size
Nearest



Instead, different size detectors generate dense boundary webs at different positions and depths along the shading gradient

3D VISION uses **Multiple-scale, depth-selective boundary webs**



Instead, **different size detectors** generate dense boundary webs at different positions and **depths** along the shading gradient

Each boundary web captures the gray shading in small compartments at its positions and depths

We SEE this pattern of shading across ALL the depths

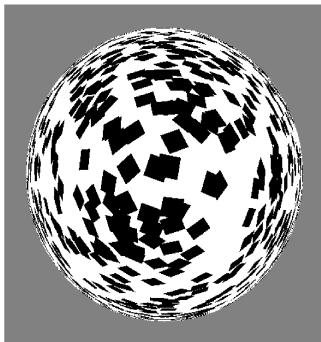
A shaded percept in depth results

Property of Stephen Grossberg.
For personal use only.

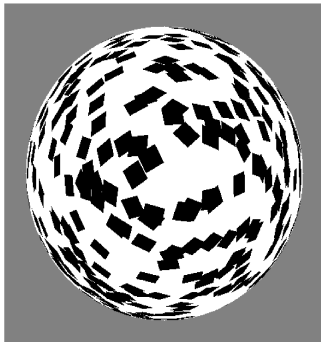
WHY SHOULD YOU BELIEVE THIS?

It explains hard data about shaded and textured surfaces

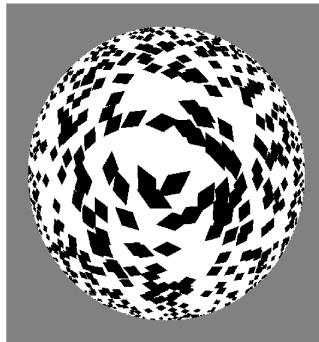
DEPTHFUL.....FLAT



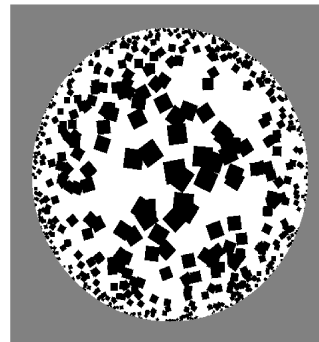
HP



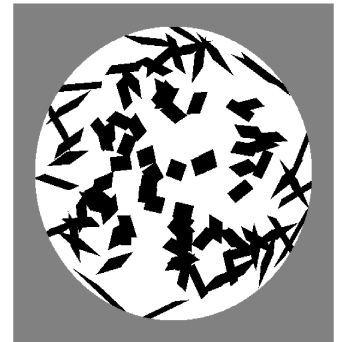
LP



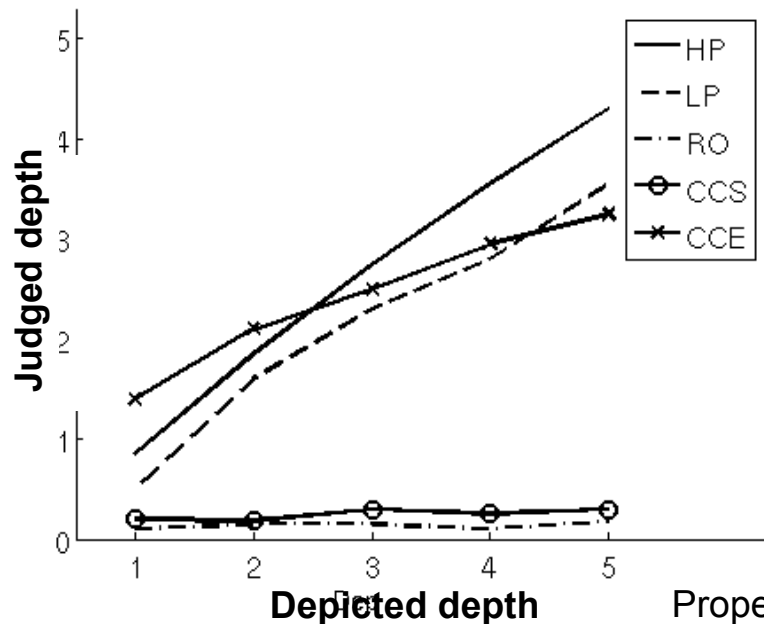
CCE



CCS



RO



Depthfulness varies with texture element width, but only when elements are elongated and sufficiently aligned with one another so as to form coherent groupings

Data of Todd and Akerstrom, 1987;
simulated in Grossberg, Kuhlmann,
and Mingolla, 2007

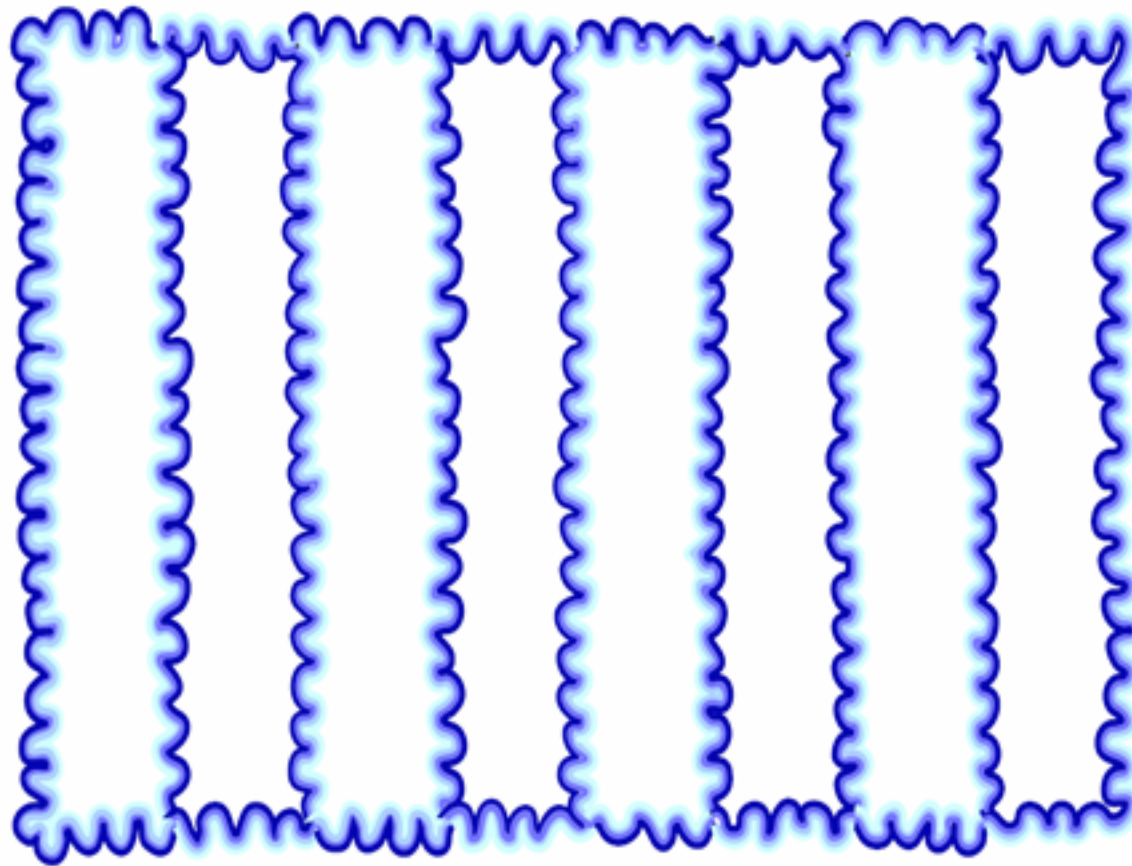
Property of Stephen Grossberg
For personal use only.

BAINGIO PINNA

Watercolor illusion, 1987

Filled-in regions **bulge in depth**

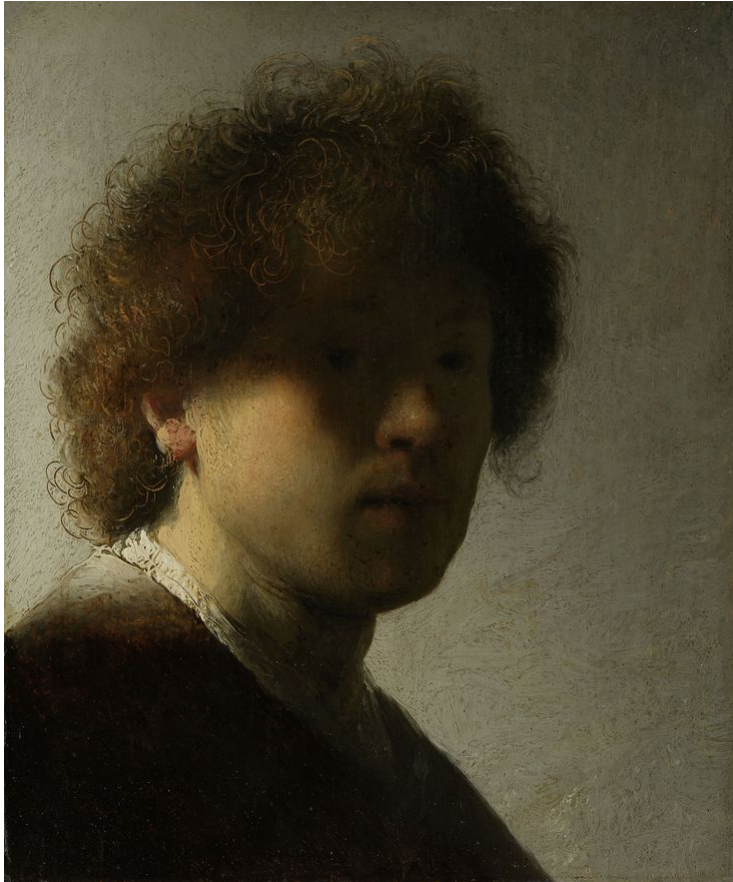
Multiple-scale, depth-selective boundary web!



Property of Stephen Grossberg.
For personal use only.

CHIAROSCURO

Rembrandt self-portrait



TROMPE L'OEIL

Graham Rust



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ROUEN CATHEDRAL

Monet, 1892-1894

At sunset

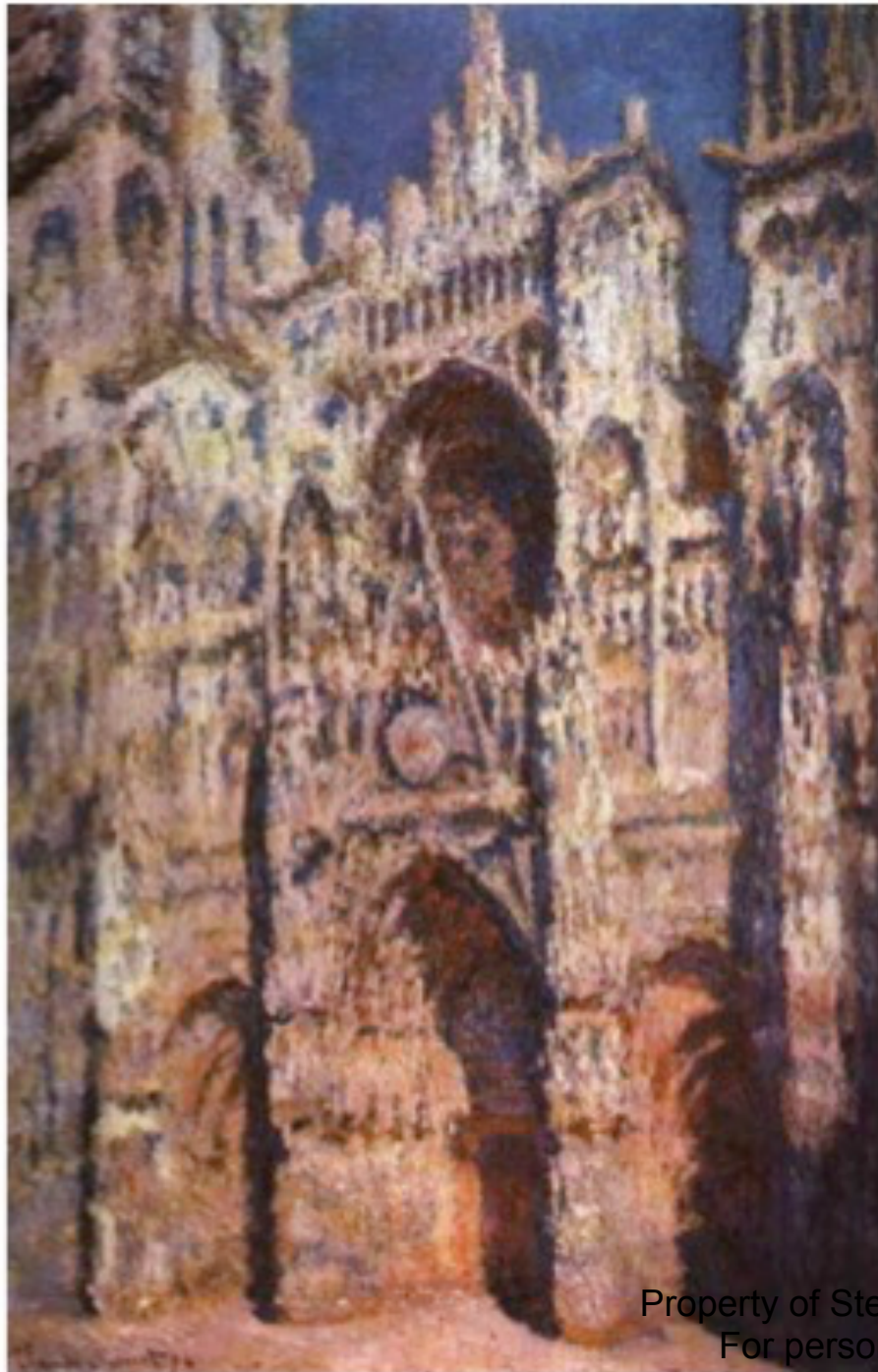
Lighting is **almost equiluminant**
across most of the painting

Here, boundaries are mostly due
to **color differences**, not
luminance differences

Fine architectural details are
obscured, leading to...

**Coarser and more uniform
boundary webs, so...**

Less depth in the painting



ROUEN CATHEDRAL

Monet, 1892-1894

Full sunlight

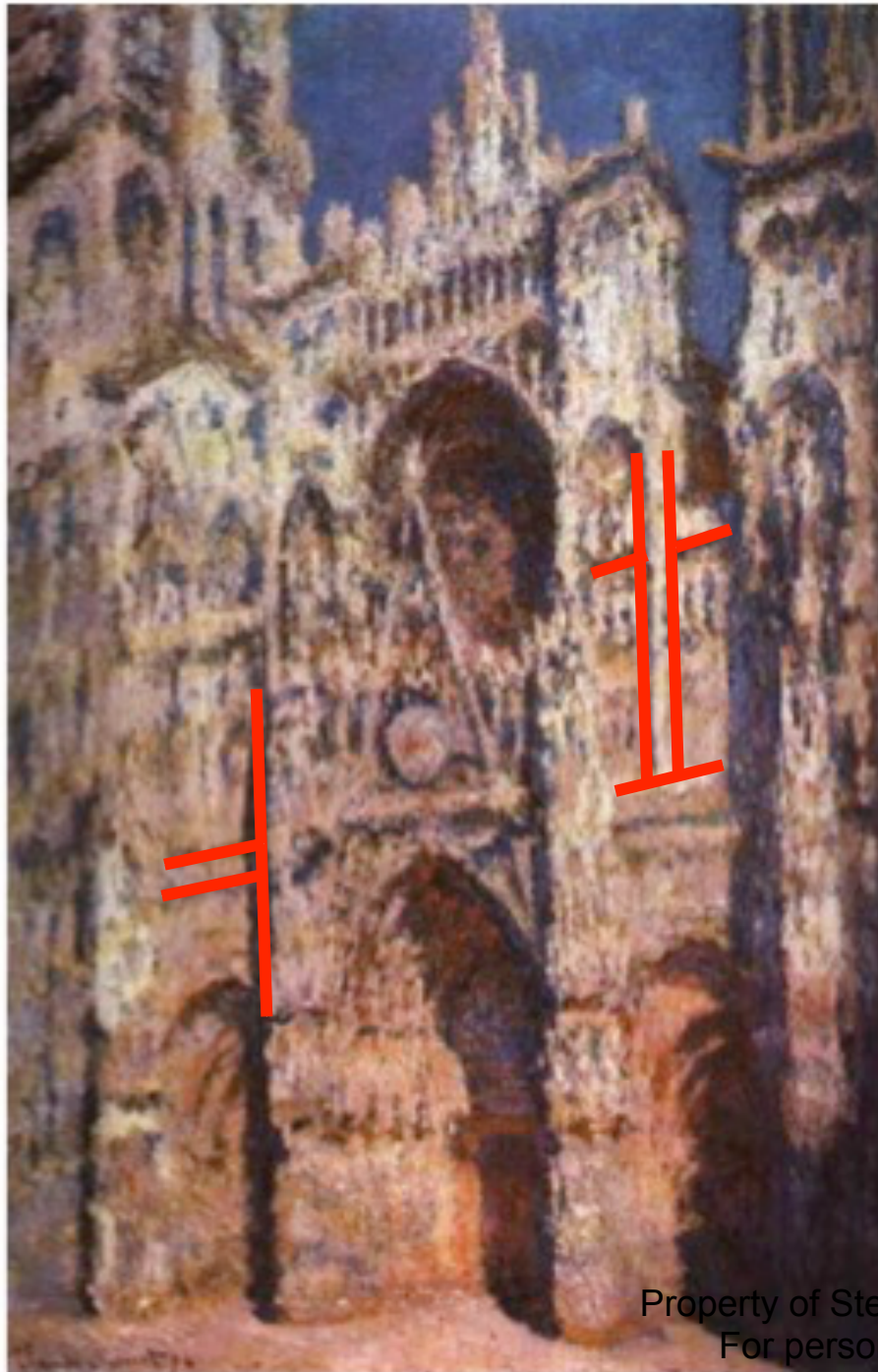
Lighting is **strongly non-uniform**
across most of the painting

Strong boundaries due to both
luminance and color differences

Fine architectural details are
much clearer, leading to...

Finer and more non-uniform
boundary webs, so...

Much more detail and depth



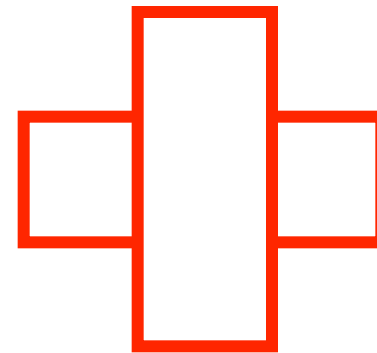
ROUEN CATHEDRAL

Monet, 1892-1894

Full sunlight

There are also more T-junctions
where vertical boundaries occlude
horizontal boundaries,
or conversely...

Leading to even more depth



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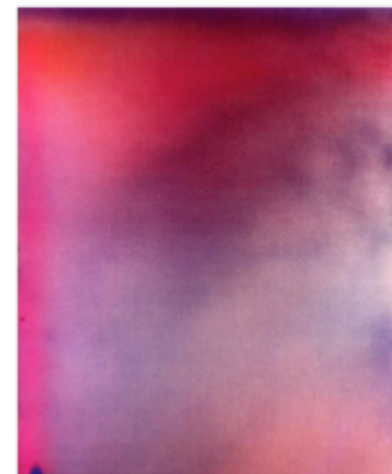
COLOR FIELD PAINTING

Jules Olitski, Spray paintings

Julius and his friends (1967)



Lysander-1 (1970)



Instant Loveland (1968) Comprehensive Dream (1965)

Property of Stephen Grossberg.
For personal use only.

COLOR FIELD PAINTING

Jules Olitski, Spray paintings

Lysander-1 (1970)



Unlike Impressionists like Monet,
no discrete colored units exist

No structured color or luminance
gradients

Boundary webs are spread over
the entire surface...

creating a percept of a space filled
with a **colored fog** and a sense of **ambiguous depth**

“When the conception of internal form is governed by edge,
color...appears to remain on or above the surface. I think...of color
as being seen in and throughout, not solely on, the surface”

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SELF-LUMINOUS PAINTINGS!

Ross Bleckner



Galaxy Painting (1993)



Galaxy with Birds (1993)

Property of Stephen Grossberg.
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SELF-LUMINOUS PAINTINGS!

Two Types of Mechanisms

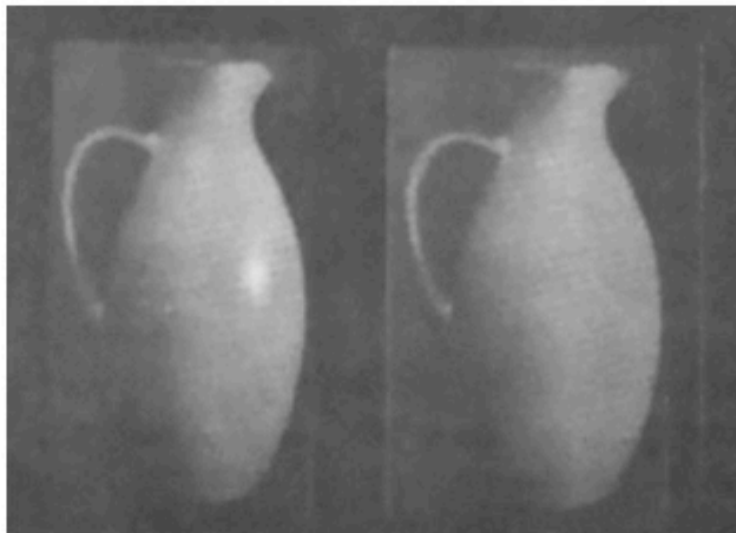
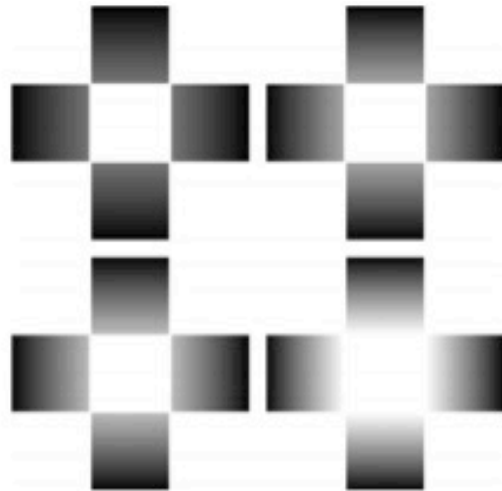
BOUNDARY WEB GRADIENTS

LIGHTNESS ANCHORING

BOUNDARY WEB GRADIENT CAN CAUSE SELF-LUMINOSITY

Similar to WATERCOLOR ILLUSION

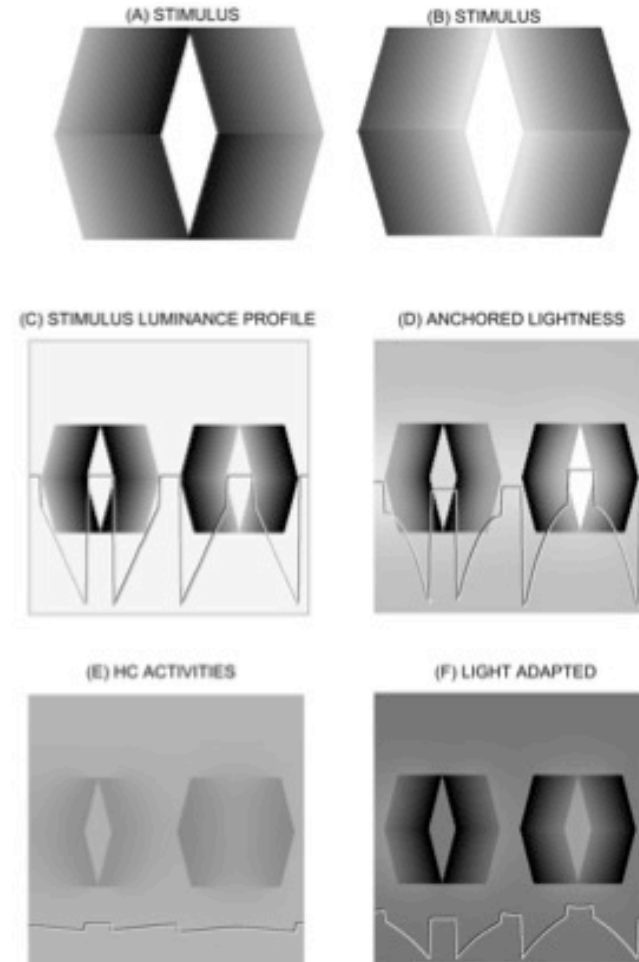
Glare



Gloss by Attached Highlight
Beck and Prazdny (1981)

Double Brilliant Illusion

Bressan (2001)



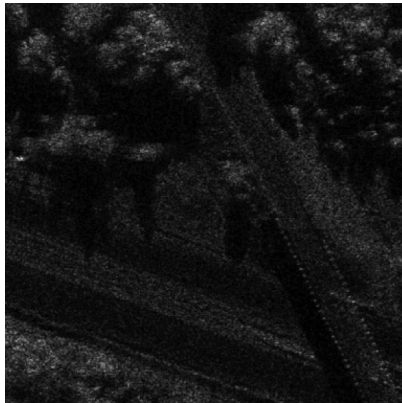
Simulation
Grossberg and Hong (2006)

Property of Stephen Grossberg.
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What is
LIGHTNESS ANCHORING?

DO THESE IDEAS WORK ON HARD PROBLEMS?

From Seurat to SAR



input



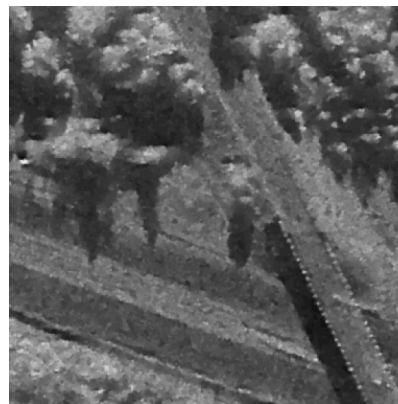
feature

Discounting the Illuminant
normalizes the image:
It preserves **RELATIVE** activities
without **SATURATION**

Still shows individual **PIXELS**



boundary



surface filling-in

Filling-in averages brightnesses
within boundary compartments

Boundaries complete between
regions where normalized
feature contrasts change



Property of Stephen Grossberg. For
personal use only.

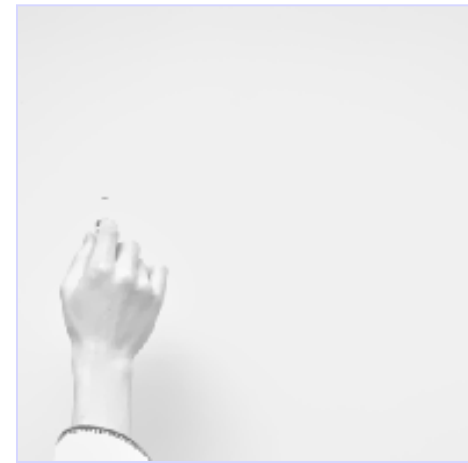
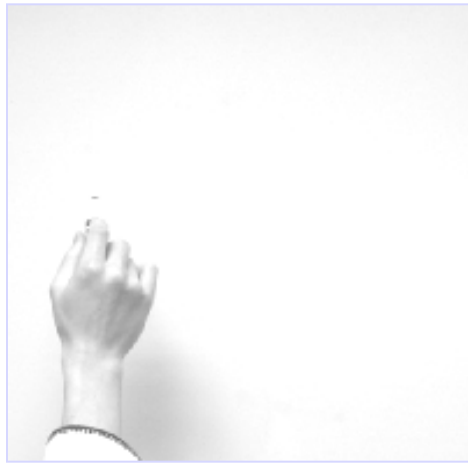
How Is the FULL DYNAMIC RANGE of a Cell Used?

How does the brain compute what is WHITE in a scene?

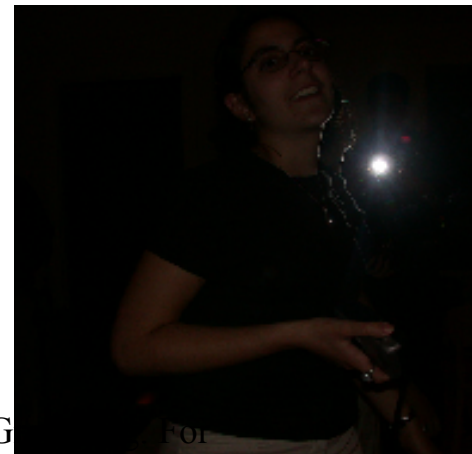
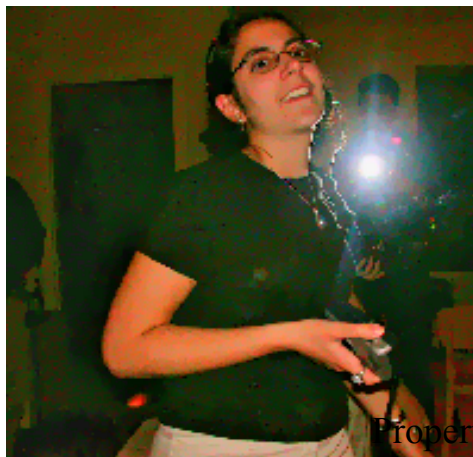
LIGHTNESS ANCHORING

HIGHEST LUMINANCE AS WHITE (HLAW) RULE

Hans Wallach (1948)



Good

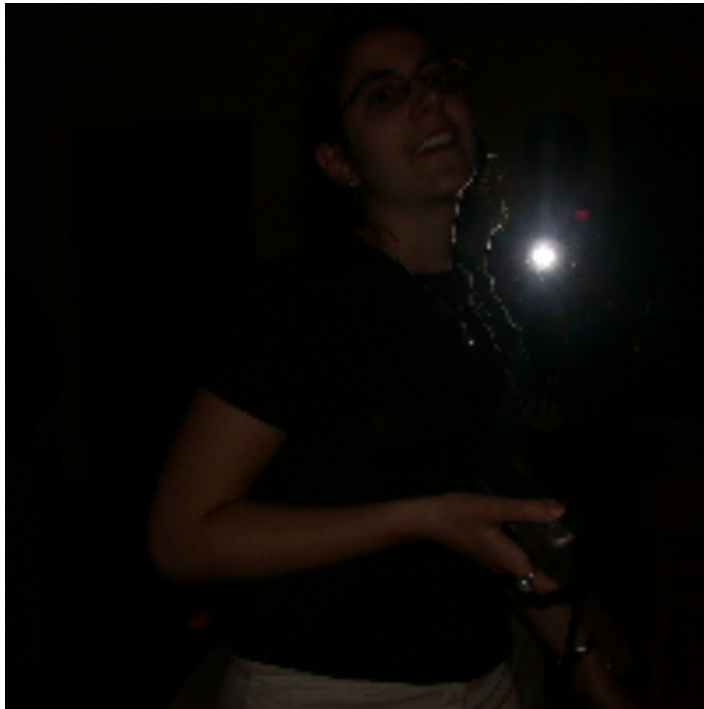


Bad!

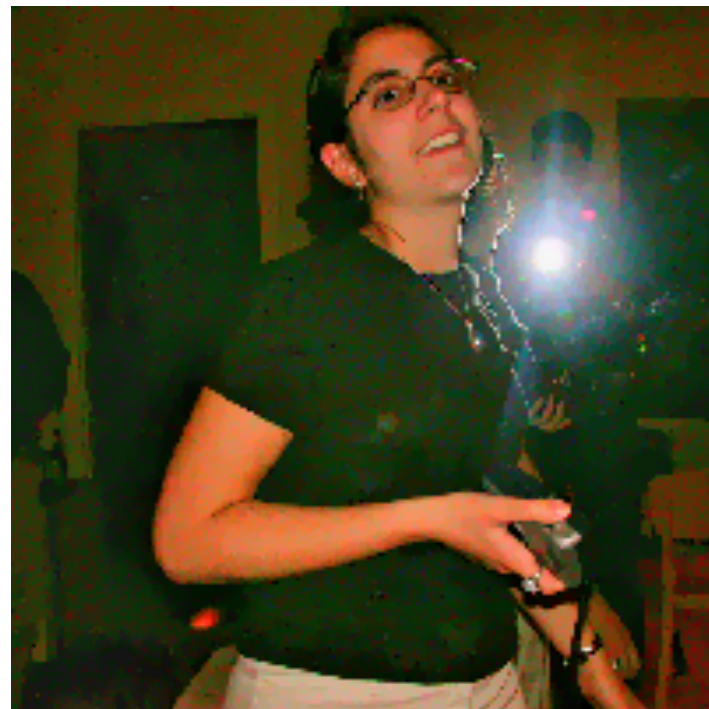
BLURRED HIGHEST LUMINANCE AS WHITE (BHLAW) RULE

Grossberg and Hong (2004, 2006)

HLAW



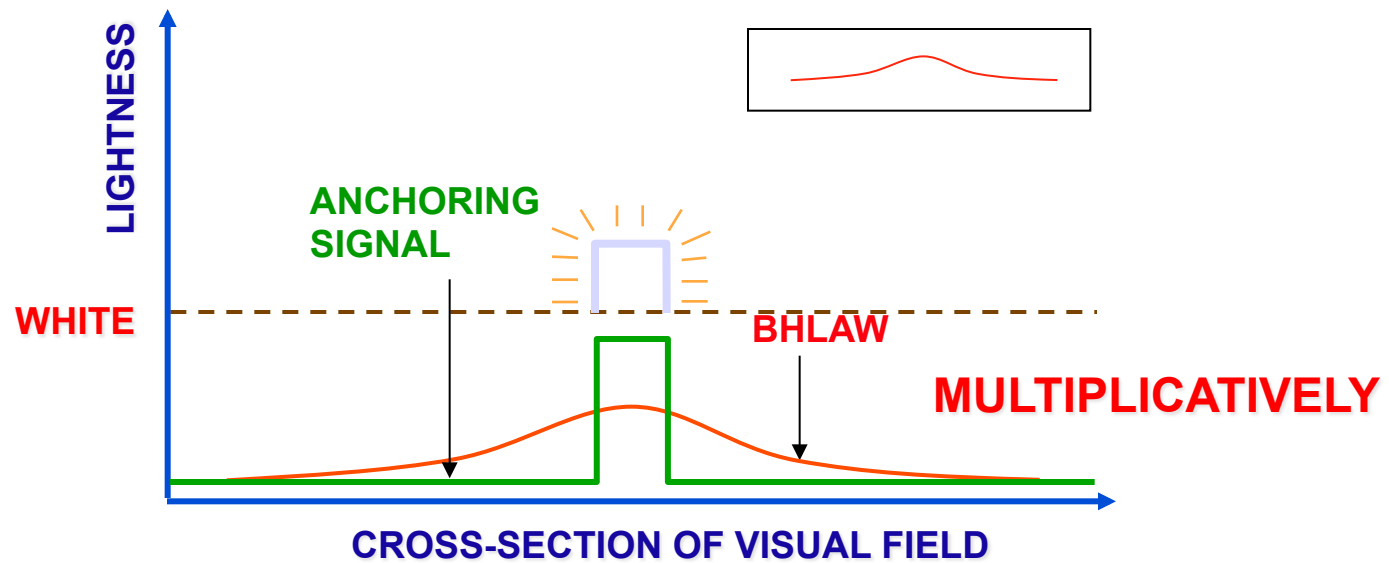
BHLAW



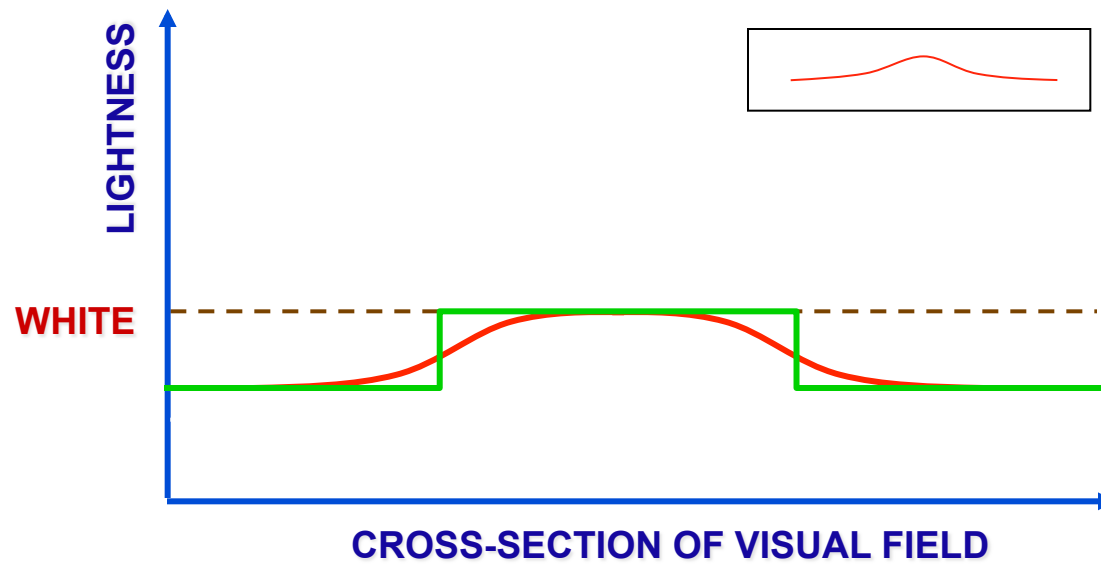
Spatial integration (blurring) adds spatial context to lightness perception

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personal use only.

BHLAW RULE



BHLAW RULE



SELF-LUMINOUS PAINTINGS!

Ross Bleckner

Anchoring



Galaxy Painting (1993)

Gradient



Galaxy with Birds (1993)

Property of Stephen Grossberg.
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HOW DO WE CONSCIOUSLY SEE A PAINTING?

...very briefly...

What is the Hard Problem of Consciousness?

Wikipedia

“...is the problem of explaining
how and why we have qualia or phenomenal
experiences...”

Chalmers (1995):

“The really hard problem of consciousness is the
problem of *experience*. When we think and perceive,
there is a whirl of information-processing,
but there is also a subjective aspect...”

What is the Hard Problem of Consciousness?

Internet Encyclopedia of Philosophy

“The hard problem of consciousness is the problem of explaining why any physical state is conscious rather than unconscious... It is the problem of explaining why...conscious mental states “light up” and directly appear to the subject.... we can still meaningfully ask the question, *Why is it conscious?...*”

Before jumping in, it is fair to ask:

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**What kind of event occurs in the brain
that is anything more than a
“whir of information processing”**

**What happens when conscious mental states
“light up”
and directly appear to a subject?**

Before jumping in, it is fair to ask:

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**Our brains sometimes go into a
context-sensitive RESONANT STATE
that can involve multiple brain regions**

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ALL CONSCIOUS STATES ARE RESONANT STATES

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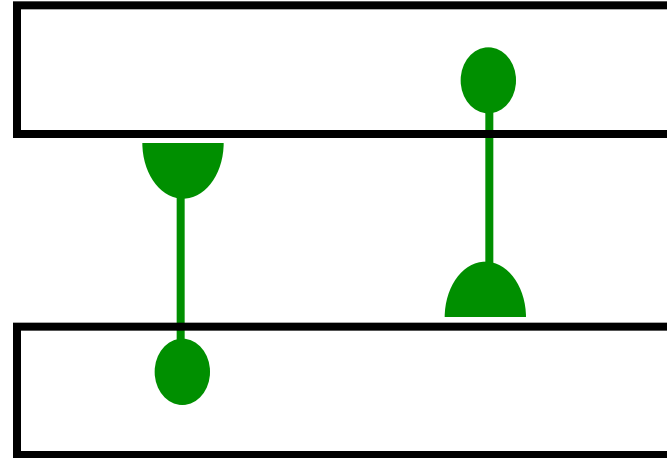
**What happens when conscious mental states
“light up”
and directly appear to a subject?**

**Our brains sometimes go into a
context-sensitive RESONANT STATE
that can involve multiple brain regions**

ALL CONSCIOUS STATES ARE RESONANT STATES
Not all brain dynamics are resonant, so
consciousness is not just a “whir of information processing”

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WHAT IS A RESONANT BRAIN STATE?



A dynamical state during which neuronal firings across a brain network are **amplified and synchronized** when they interact via **reciprocal excitatory feedback signals** during a **matching process** that occurs between **bottom-up and top-down pathways**

CENTRAL CLAIM

Conscious states are part of larger adaptive behavioral capabilities that help us to adapt to a changing world

Resonances for conscious

seeing help to ensure effective reaching

hearing help to ensure effective speaking

feeling help to ensure effective goal-oriented action

CENTRAL CLAIM

Conscious states are part of larger adaptive behavioral capabilities that help us to **adapt to a changing world**

Resonances for conscious

seeing help to ensure effective **reaching**

hearing help to ensure effective **speaking**

feeling help to ensure effective **goal-oriented action**

WHY DID EVOLUTION INVENT CONSCIOUSNESS?

**Visual inputs to the retina are
ambiguous, noisy, and incomplete**

**Multiple processing stages are needed to generate
a sufficiently complete and stable surface representation
with which to control effective looking and reaching**

A SURFACE-SHROUD RESONANCE
“lights up” this surface representation
with an extra degree of freedom
CONSCIOUS AWARENESS!
and lets the brain use IT to control
looking and reaching

CLASSIFICATION OF RESONANCES

Surface-shroud resonances support conscious seeing
of visual qualia **SEEING**

Feature-category resonances support conscious recognition
of visual objects and scenes **KNOWING**

Stream-shroud resonances support conscious hearing
of auditory qualia

Spectral-pitch-and-timbre resonances support conscious
recognition of sources in auditory streams

Item-list resonances support conscious recognition of
speech and language

Cognitive-emotional resonances support conscious feelings
and recognition of them

ADAPTIVE RESONANCE THEORY

ART

Grossberg (1976)

A unifying theme:

Stability-Plasticity Dilemma

How can learning continue into adulthood without causing catastrophic forgetting?

How can we LEARN quickly without being forced to FORGET just as quickly?

e.g., why learning your faces does not force me to forget faces of my family and friends!

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ART has been incrementally developed ever since

**ART is currently the most advanced cognitive and neural
theory...with the broadest explanatory and predictive range...
about how brains learn to**

attend

recognize

predict

objects and events in a changing world

**ALL of the main ART predictions have been supported by
psychological and neurobiological data**

ART WORKS!

LARGE-SCALE APPLICATIONS TO ENGINEERING AND TECHNOLOGY

<http://www.cns.bu.edu/techlab> (Gail Carpenter)

Boeing parts design retrieval (used to design Boeing 777)

satellite remote sensing

radar identification

robot sensory-motor control and navigation

machine vision

3D object and face recognition

Macintosh operating system software

automatic target recognition

ECG wave recognition

protein secondary structure identification

character classification

musical analysis

air quality monitoring and weather prediction

medical imaging and database analysis

multi-sensor chemical analysis

strength prediction for concrete mixes

signature verification

decision making and intelligent agents

machine condition monitoring and failure forecasting

chemical analysis

electromagnetic and digital circuit design

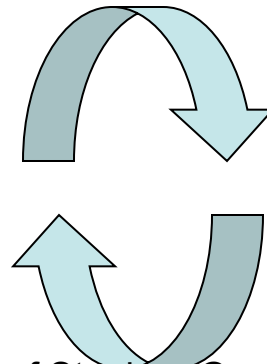
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ART MAIN IDEA

Top-down attentive feedback
encodes
learned expectations
that
STABILIZE LEARNING AND MEMORY
in response to
a changing world
that is filled with
unexpected events

**Attentive Information
Processing**

FAST



**Learning and
Memory**

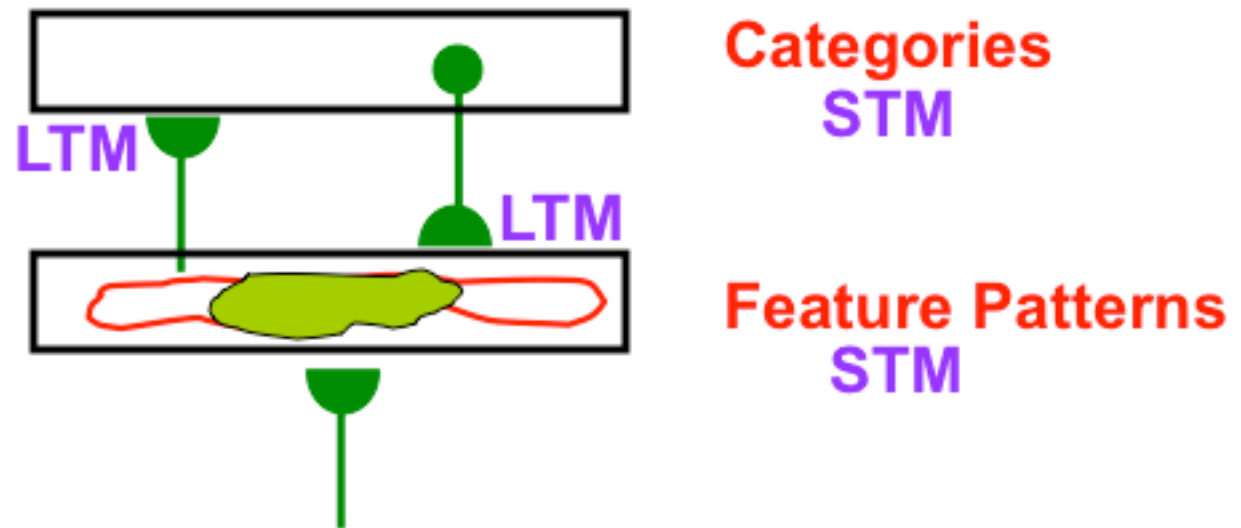
SLOW

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ADAPTIVE RESONANCE

Attended feature clusters reactivate bottom-up pathways

Activated categories reactivate their top-down pathways



Feature-Category resonance synchronizes
amplifies
prolongs system response

Resonance triggers learning in bottom-up and top-down
adaptive weights: adaptive resonance!

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CLASSIFICATION OF RESONANCES

Surface-shroud resonances support conscious seeing
of visual qualia

Feature-category resonances support conscious recognition
of visual objects and scenes

Stream-shroud resonances support conscious hearing
of auditory qualia

Spectral-pitch-and-timbre resonances support conscious
recognition of sources in auditory streams

Item-list resonances support conscious recognition of
speech and language

Cognitive-emotional resonances support conscious feelings
and recognition of them

WHAT IS AN ATTENTIONAL SHROUD?

Surface-fitting spatial attention
ATTENTIONAL SHROUD!
marks the object-hood of the
as-yet-undefined object category

Tyler and Kontsevich (1995)
used shrouds to study
perceptual transparency

Cf. Cavanagh, Pylyshyn, Yantis,...



Magritte (1927)

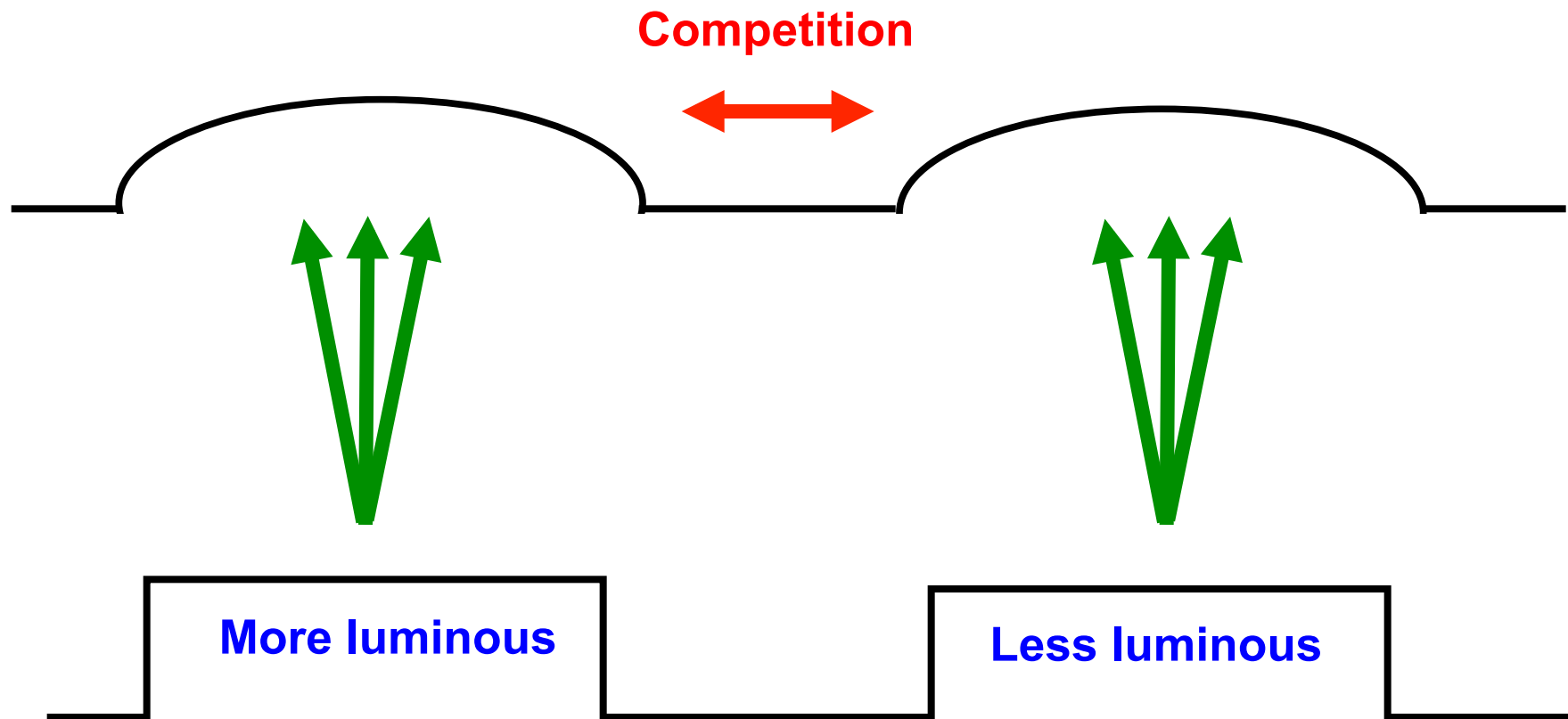
PREDICTION:
Shrouds enable learning of view-invariant object categories

Grossberg (2007, 2009)
Fazl, Grossberg, and Mingolla (2009)
Cao, Grossberg, and Markowitz (2011)
Grossberg, Markowitz, and Cao (2011)
Foley, Grossberg, and Mingolla (2012)
Chang, Grossberg, and Cao (2014)

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BOTTOM-UP SPATIAL ATTENTIONAL COMPETITION

Spatial Attention

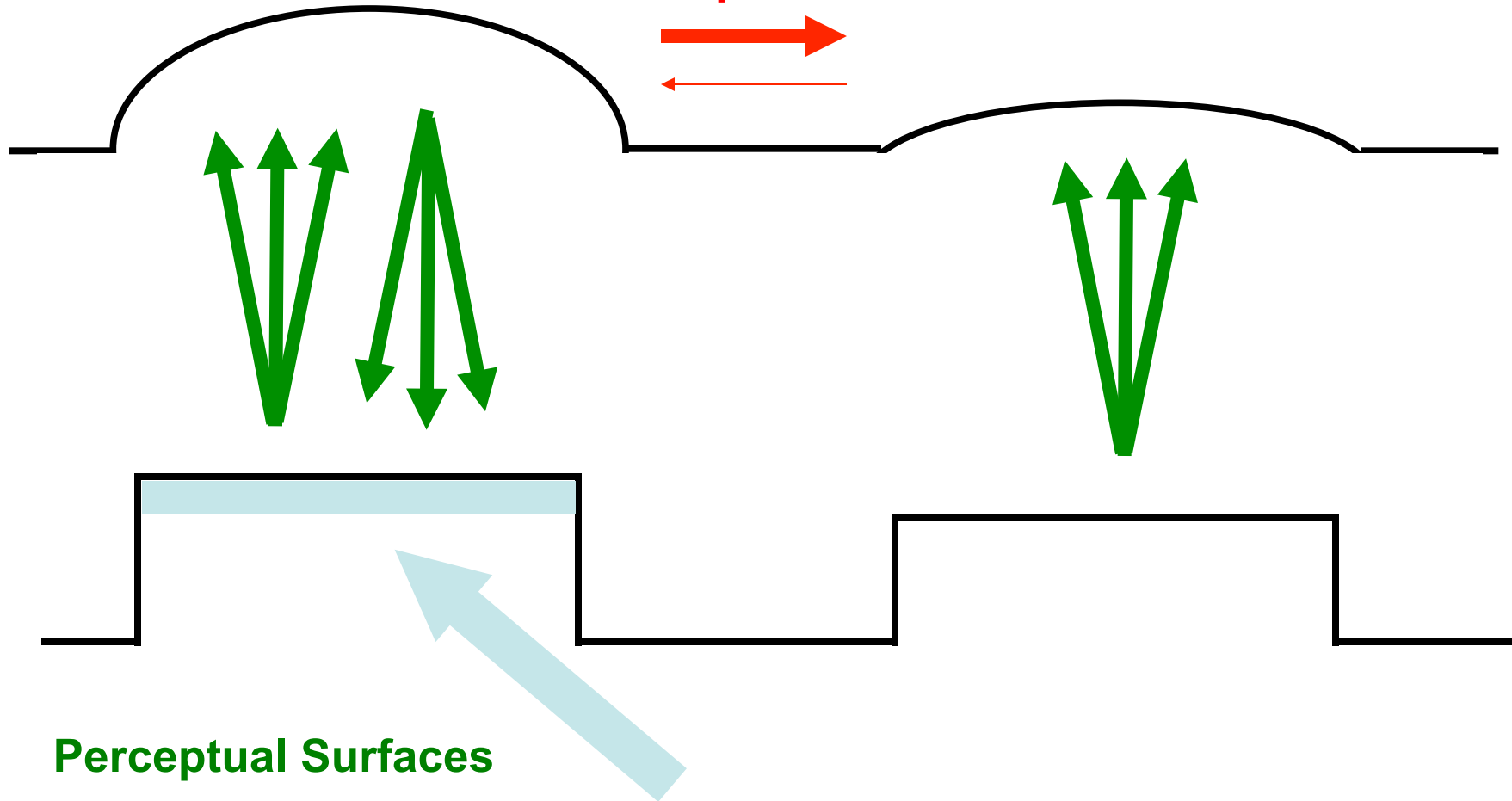


Perceptual Surfaces

SURFACE-SHROUD RESONANCE

Spatial Attention

Competition



Perceptual Surfaces

Psychophysics: Carrasco, Penpeci-Talgar, and Eckstein (2000)
Neurophysiology: Reynolds and Desimone (2003)

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SURFACE-SHROUD RESONANCE

An active
SURFACE-SHROUD RESONANCE
means that sustained
SPATIAL ATTENTION IS FOCUSED
ON THE OBJECT SURFACE

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A SURFACE-SHROUD RESONANCE ALSO SUPPORTS
CONSCIOUS SEEING OF
AN ATTENDED OBJECT

THAT IS HOW WE CONSCIOUSLY SEE A PAINTING!

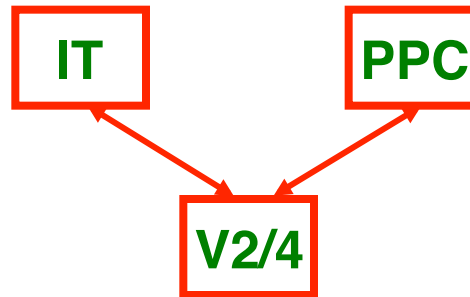
See my 2017 article on the Hard Problem of Consciousness!

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WHAT KINDS OF RESONANCES SUPPORT KNOWING VS. SEEING?

What Stream

Where Stream



KNOWING

Feature-Category
Resonance

SEEING

Surface-Shroud
Resonance

Synchronous linkage between resonances enables us to
KNOW what the object is as we **SEE** it

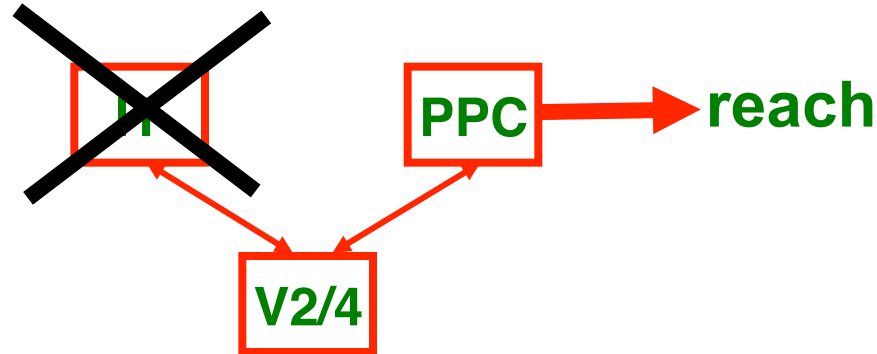
THAT'S A MONET!!!

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WHAT KINDS OF RESONANCES SUPPORT KNOWING VS. SEEING?

What Stream

Where Stream



KNOWING

Feature-Category
Resonance

SEEING

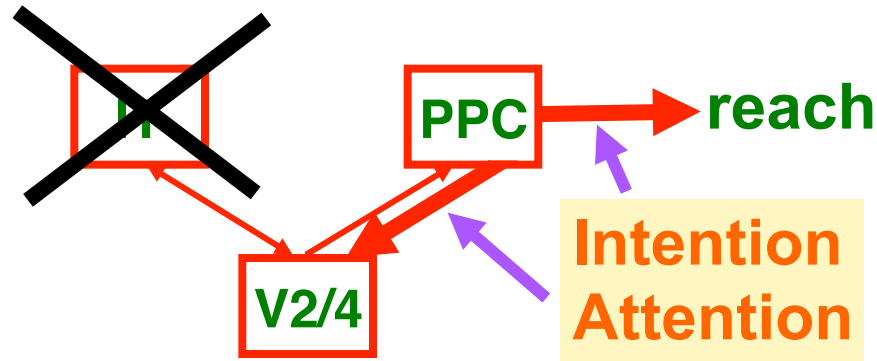
Surface-Shroud
Resonance

VISUAL AGNOSIA: reaching without knowing
Patient DF Goodale et al, 1991

WHAT KINDS OF RESONANCES SUPPORT KNOWING VS. SEEING?

What Stream

Where Stream



KNOWING

Feature-Category
Resonance

SEEING

Surface-Shroud
Resonance

VISUAL AGNOSIA: reaching without knowing
Patient DF Goodale et al, 1991

Attention and Intention both parietal cortical functions
Andersen, Essick, and Siegel, 1985; Gnadt and Andersen, 1988; Snyder, Batista, and Andersen, 1997, 1998

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HOW WE SEE ART AND HOW ARTISTS MAKE IT

I've sketched some of the brain designs
that help to understand this:

Complementary computing of boundaries and surfaces
Hierarchical resolution of uncertainty
Multiple-scale boundary webs
Lightness anchoring
Surface-shroud resonance

You can begin to see how our brains compute
VERY DIFFERENTLY from traditional computers!
VERY DIFFERENTLY from...e.g....Deep Learning!

I did NOT mention **BEAUTY** or how we **FEEL** about a painting

For that, you need to study how
COGNITIVE-EMOTIONAL
processes occur

a discussion for another day!

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