



TSBB15

Computer Vision

Lecture 9

Biological Vision Systems



Two parts

1. Systems perspective

2. Visual perception



Two parts

1. Systems perspective

Based on Michael Land's and Dan-Eric Nilsson's work



2. Visual perception

Based on Slides from Gösta Granlund





Vision Systems



Camera vs. eye



Vision Systems



≠



Purpose:

Reproduce the world
as accurately as possible

Purpose:

Sensing device for
visual behaviours



Vision Systems



What a camera sees



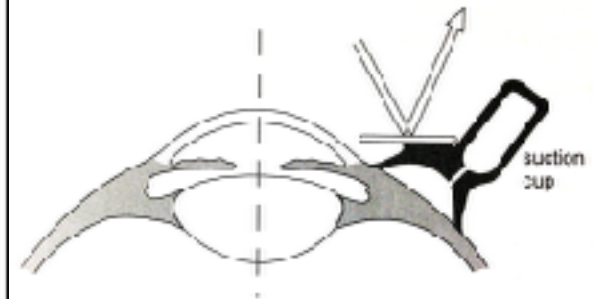
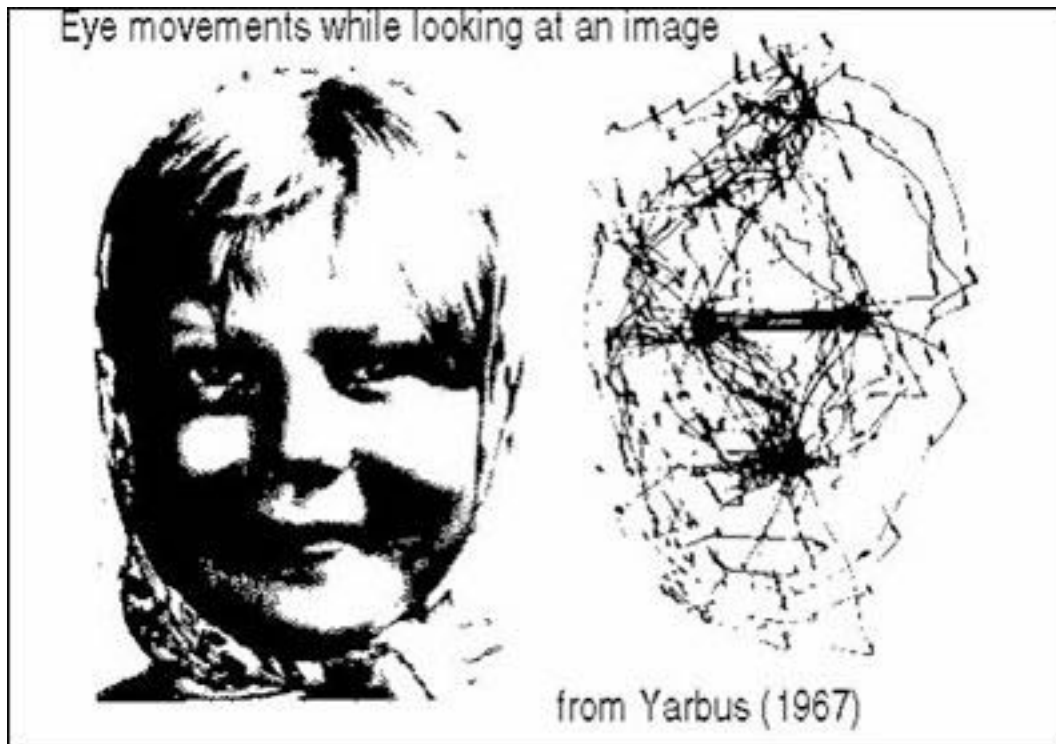
Vision Systems



What the human eye sees



Vision Systems



Device used by Yarbus
Illustration from:
M. F. Land "Looking and Acting"



Vision Systems



Uniform resolution
Smooth motion

≠



just central 2° are sharp
saccadic motions (avg. 3Hz,
around 700°/s)



Vision Systems

Peripheral view



Foveal view



What a robot sees



Visual Behaviours

Saccadic motion is an example of a visual behaviour

Purpose?



Visual Behaviours

Other examples of visual behaviours:

- 1. Fixate moving targets**
- 2. Compensate for head and body movement**
- 3. Change detection**
- 4. Recognition**



Visual Behaviours

Other examples of visual behaviours:

- 1. Fixate moving targets - Optokinetic Reflex (OKR)**
- 2. Compensate for head and body movement**
- Vestibulo - Ocular Reflex (VOR)
- 3. Change detection**
- 4. Recognition**



Visual Behaviours

Experiment:

Hold out your hand and raise a finger:

1. turn head while looking at finger (VOR)
2. move hand while looking at finger (OKR)

Which reflex is faster?



Visual Behaviours

Visual input for VOR (stabilization)?

Visual input for OKR (tracking)?



Visual Behaviours

Visual input for VOR (stabilization)?

- Optical flow (dense over entire visual field)

Visual input for OKR (tracking)?

- Tracking (region around fovea)



Visual Behaviours

Visual input for VOR (stabilization)?

- Optical flow (dense over entire visual field)

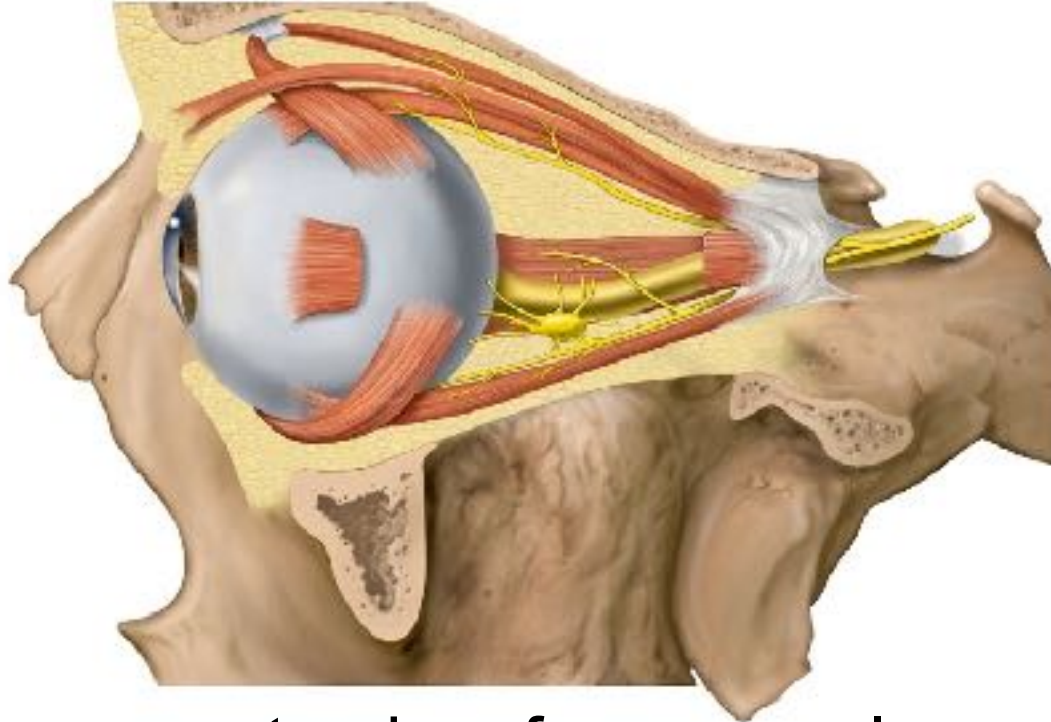
Visual input for OKR (tracking)?

- Tracking (region around fovea)

Note: VOR mainly uses input from the vestibular system (optical flow is used for learning).



Visual Behaviours



- Three opponent pairs of eye muscles
- Whole neck-eye system is involved in gaze control



VCR in Weka bird

Whole head has to move in birds - Vestibulo-Collic Reflex



[Weka VCR - YouTube](#)



VCR in Chicken

Whole head has to move in birds - Vestibulo-Collic Reflex

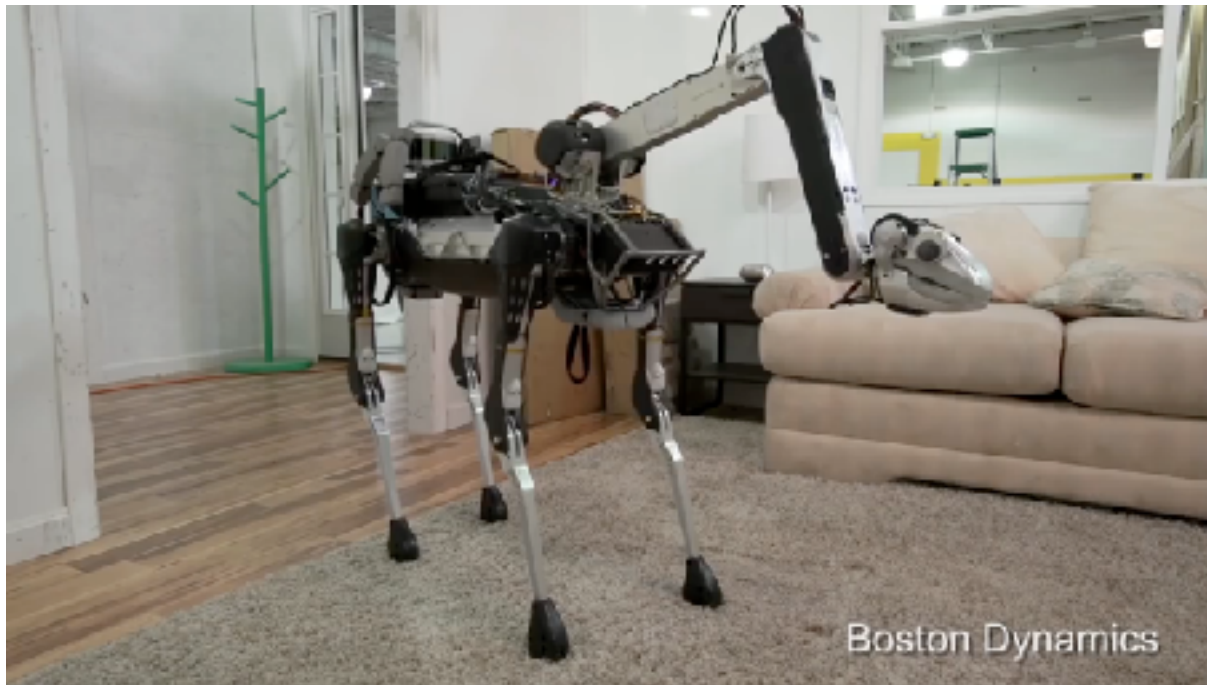


[Chicken VCR - YouTube](#)



VCR on Robot

Boston Dynamics version of VCR



[Boston Dynamics - YouTube](#)



Visual Behaviours

Examples of visual behaviours:

1. **Fixate moving targets** - OKR
2. **Compensate for head and body movement**
- VOR, VCR
3. **Change detection** - 1&2 + time difference
4. **Recognition** - Saccadic motions + 1&2 +
Perceptual hierarchy

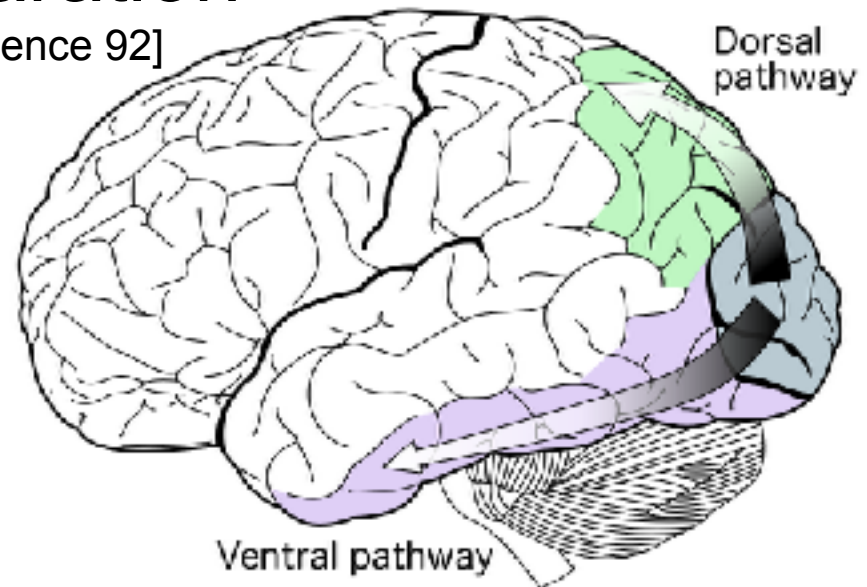


Visual Perception

- How and what separation

[Godale & Milner, Trends Neuroscience 92]

- **Dorsal pathway** controls gaze and action
- **Ventral pathway** handles visual recognition





Complex problem

Recognition using direct matching to prototype images is untenable:

- Large number of objects
- Large number of variations





Complex problem

Recognition using direct matching to prototype images is untenable:

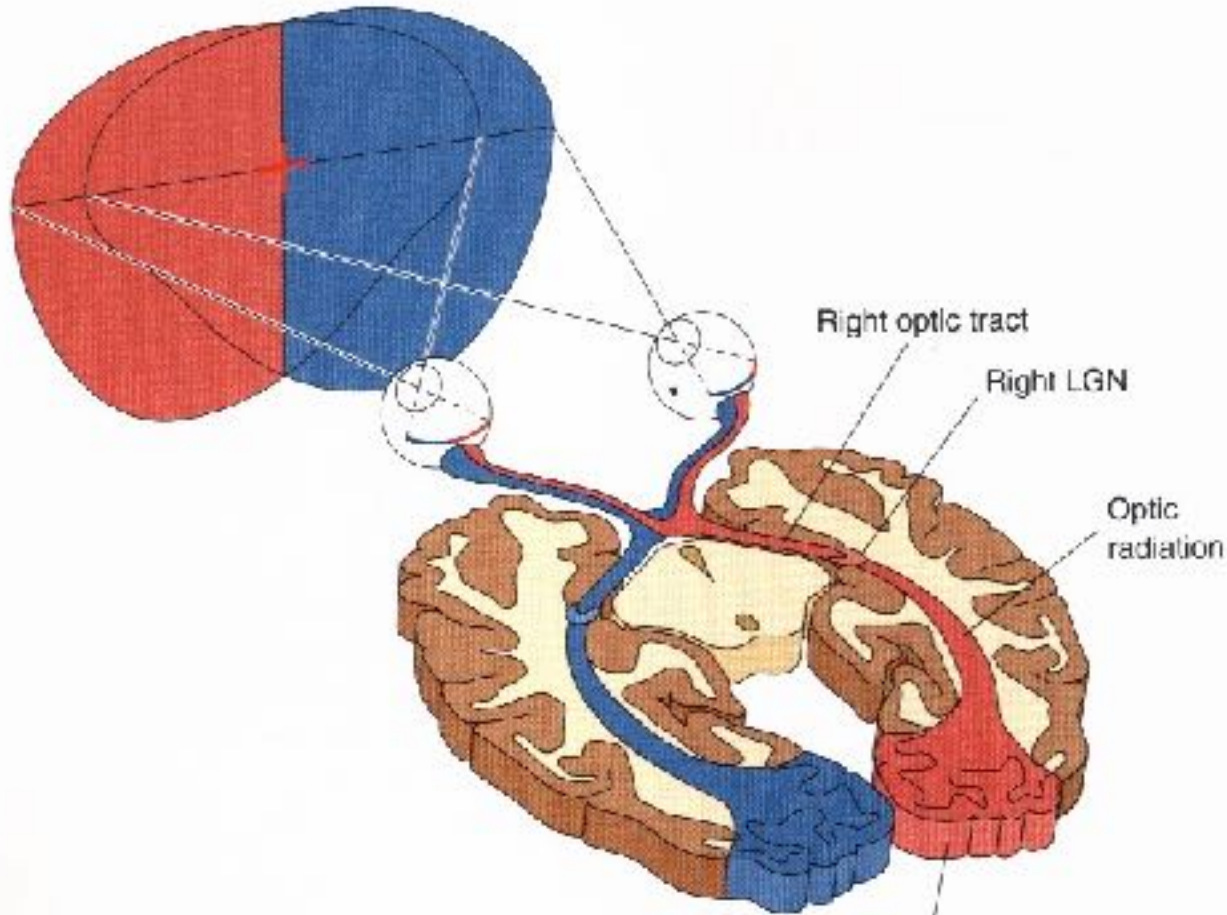
- Large number of objects
- Large number of variations



Abstraction is necessary!

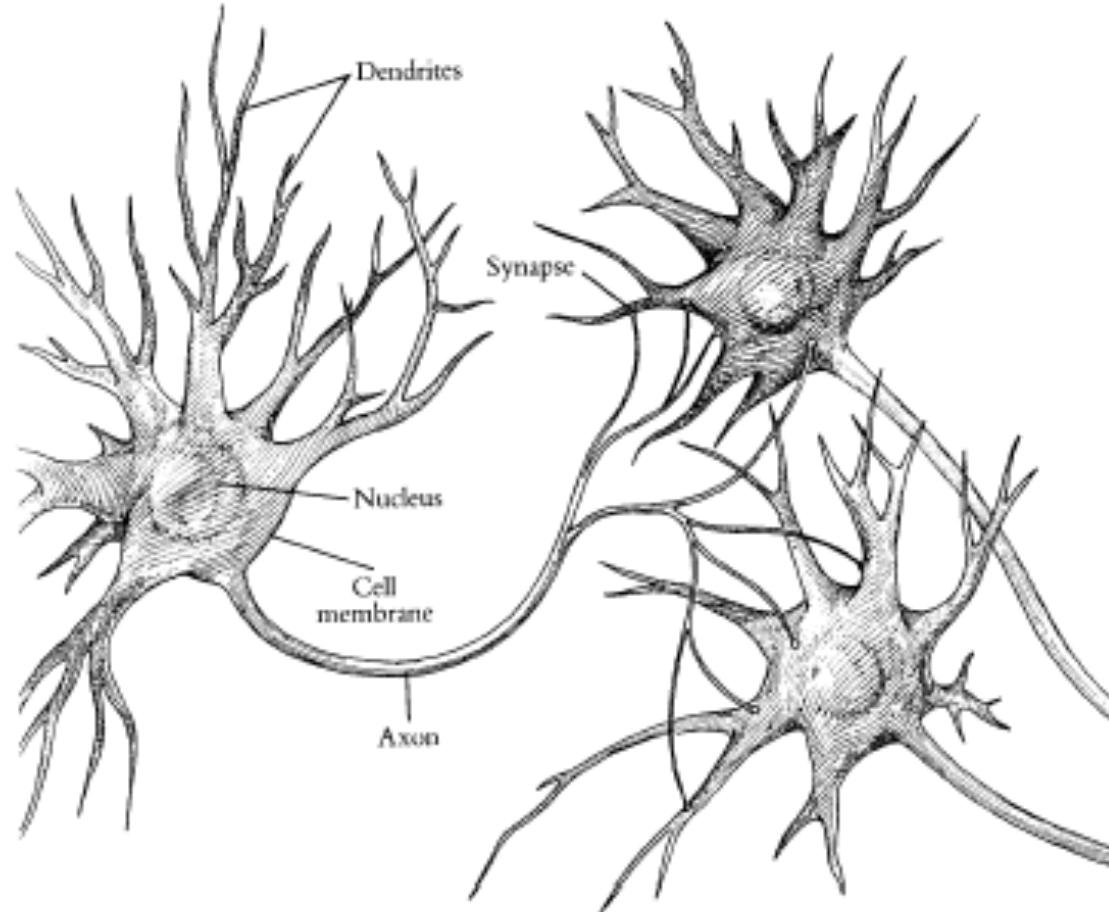


The visual pathway





Principal parts of a nerve cell





Signals of neurons

- Carried through a chemical process
- Resting potential -70 mV inside axon
- Reversal to $+40$ mV inside axon
- Refractory time about 1 msec
- A few to > 1000 impulses per second
- Most neurons use **pulse frequency coding**
- A few types have **graded** signals



Neurons

Axons can be < 1 mm to > 1 m

Synapses can be excitatory or inhibitory

50 – 100 neurotransmitters

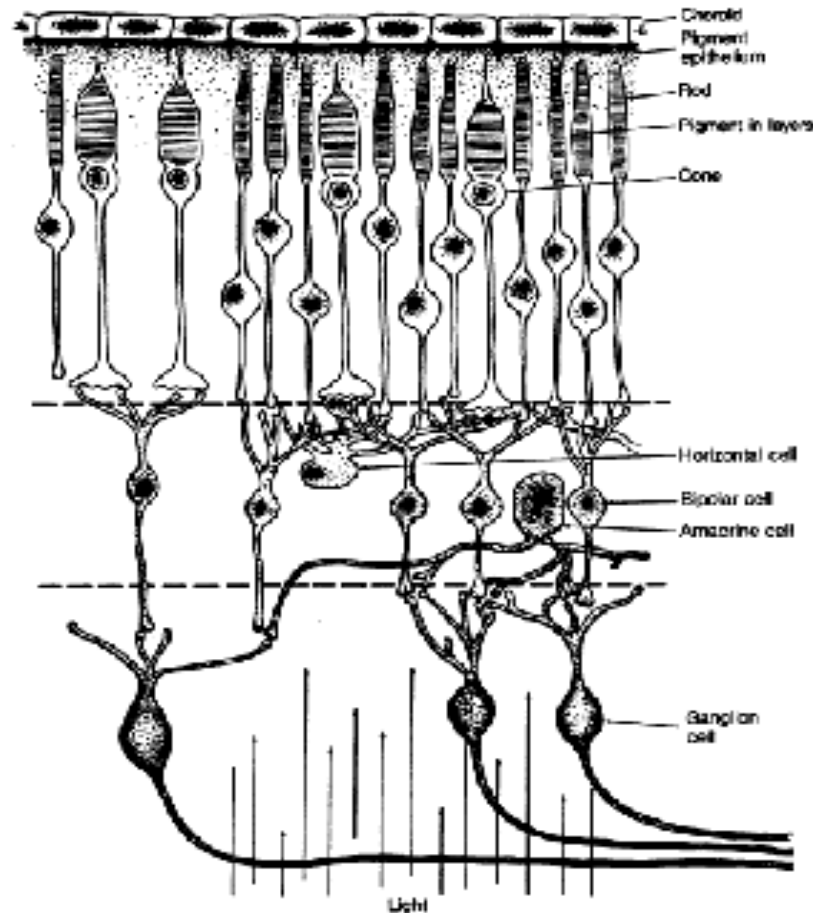


> 100 different types of neurons





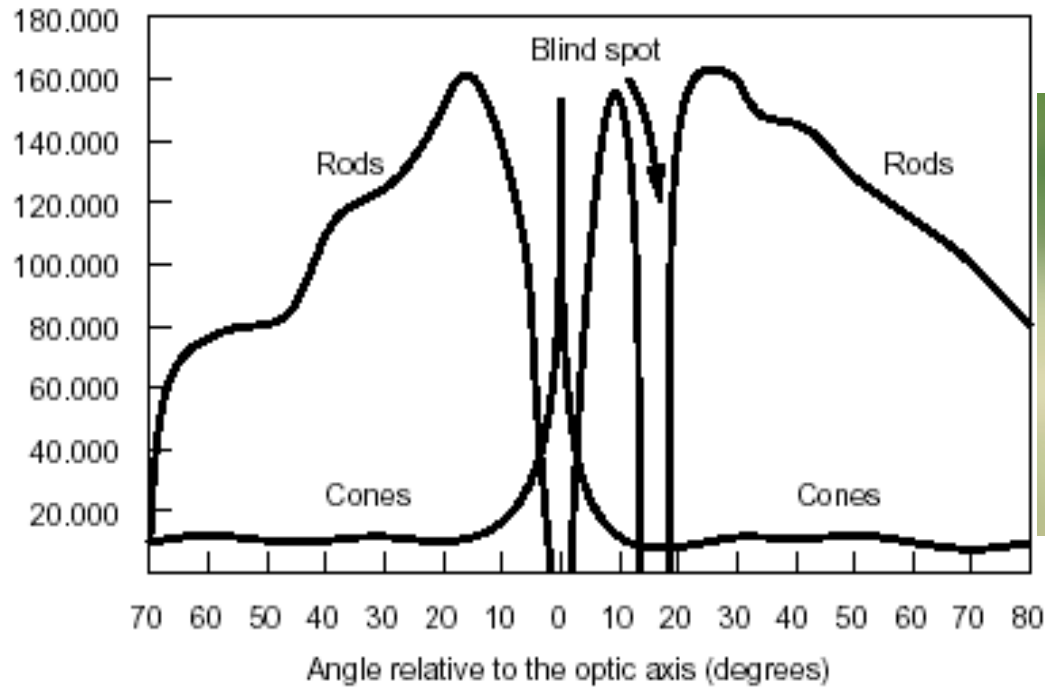
The retina





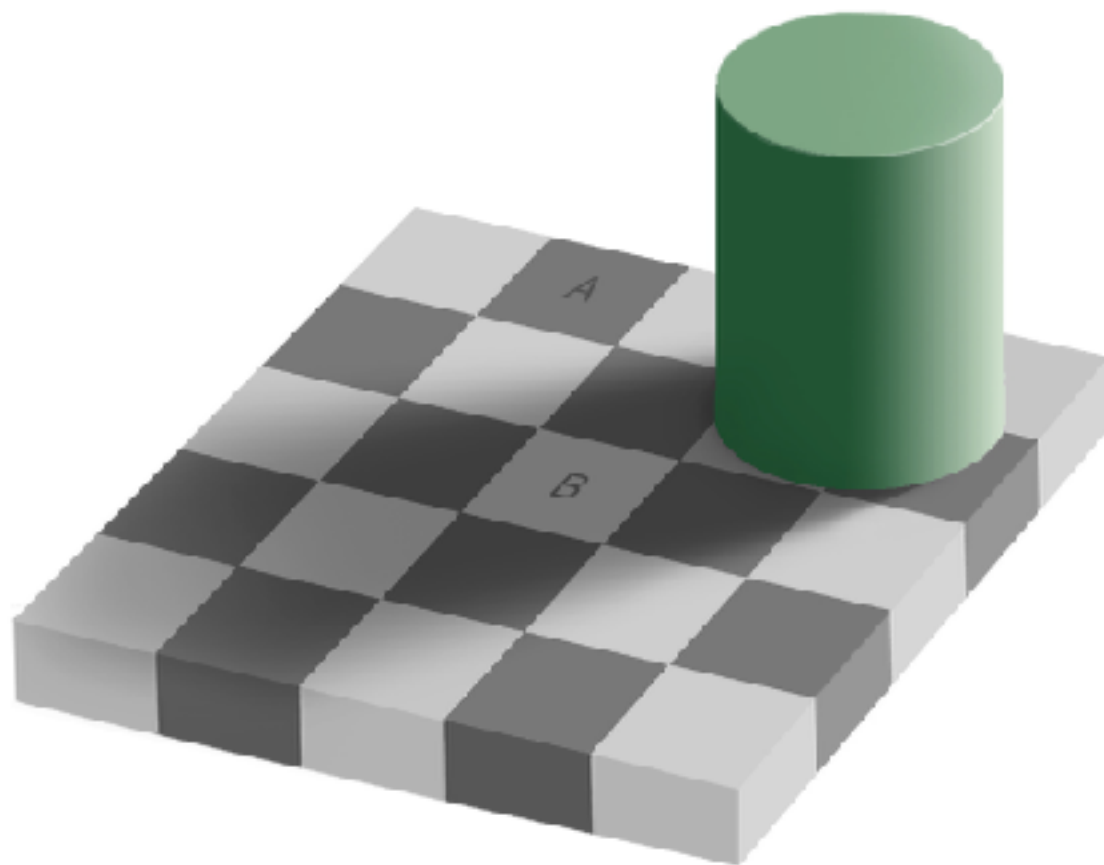
Density of photoreceptors

Number of photoreceptors
per square millimeter



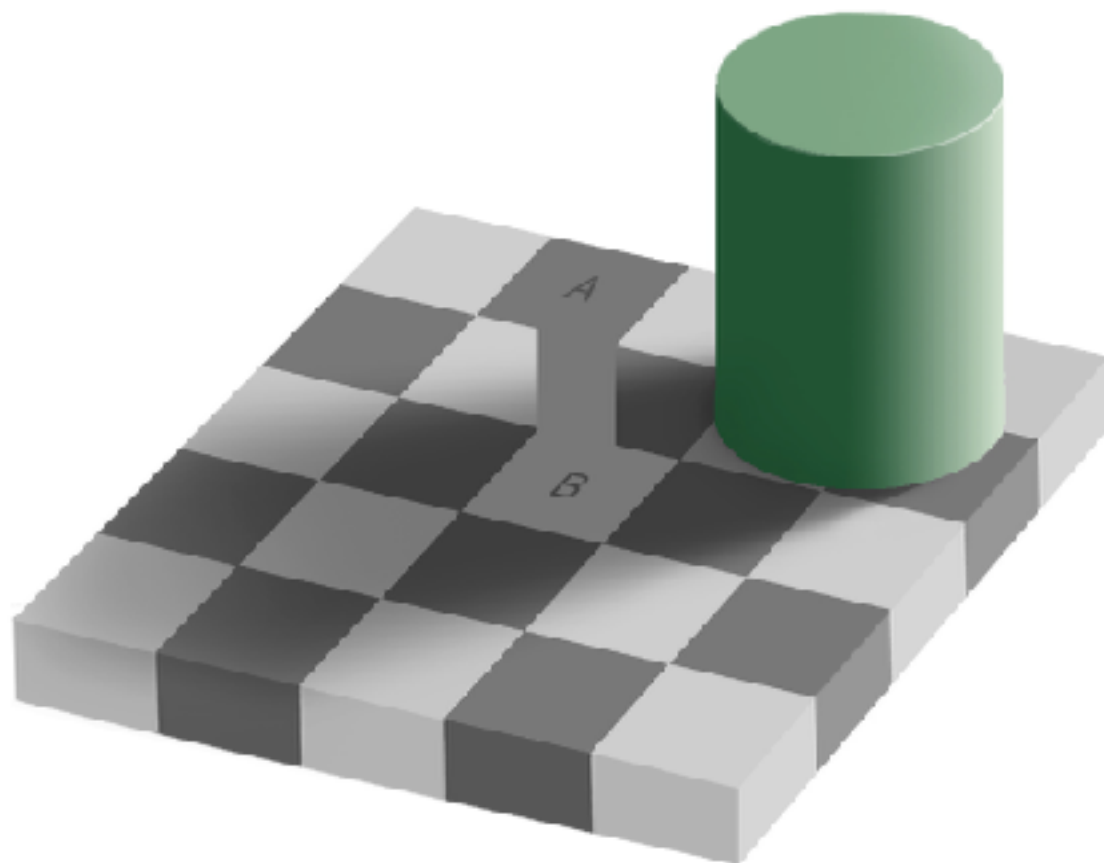


Stability with respect to illumination



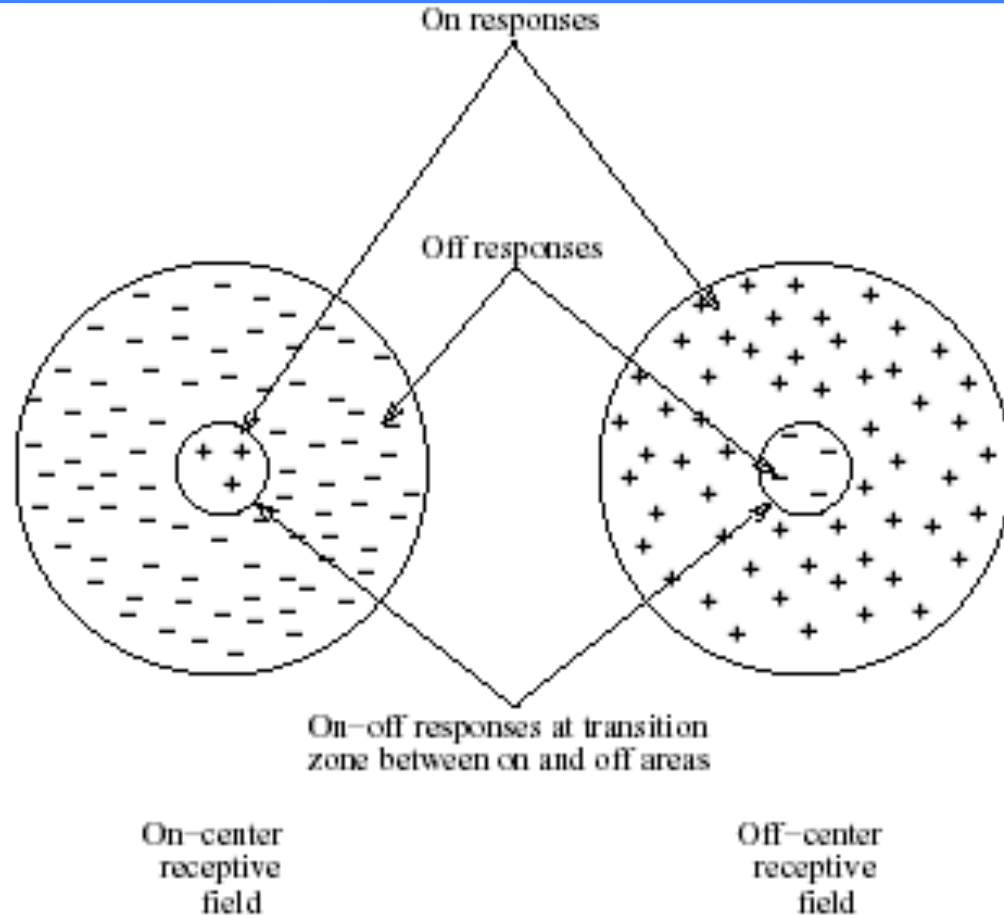


Stability with respect to illumination



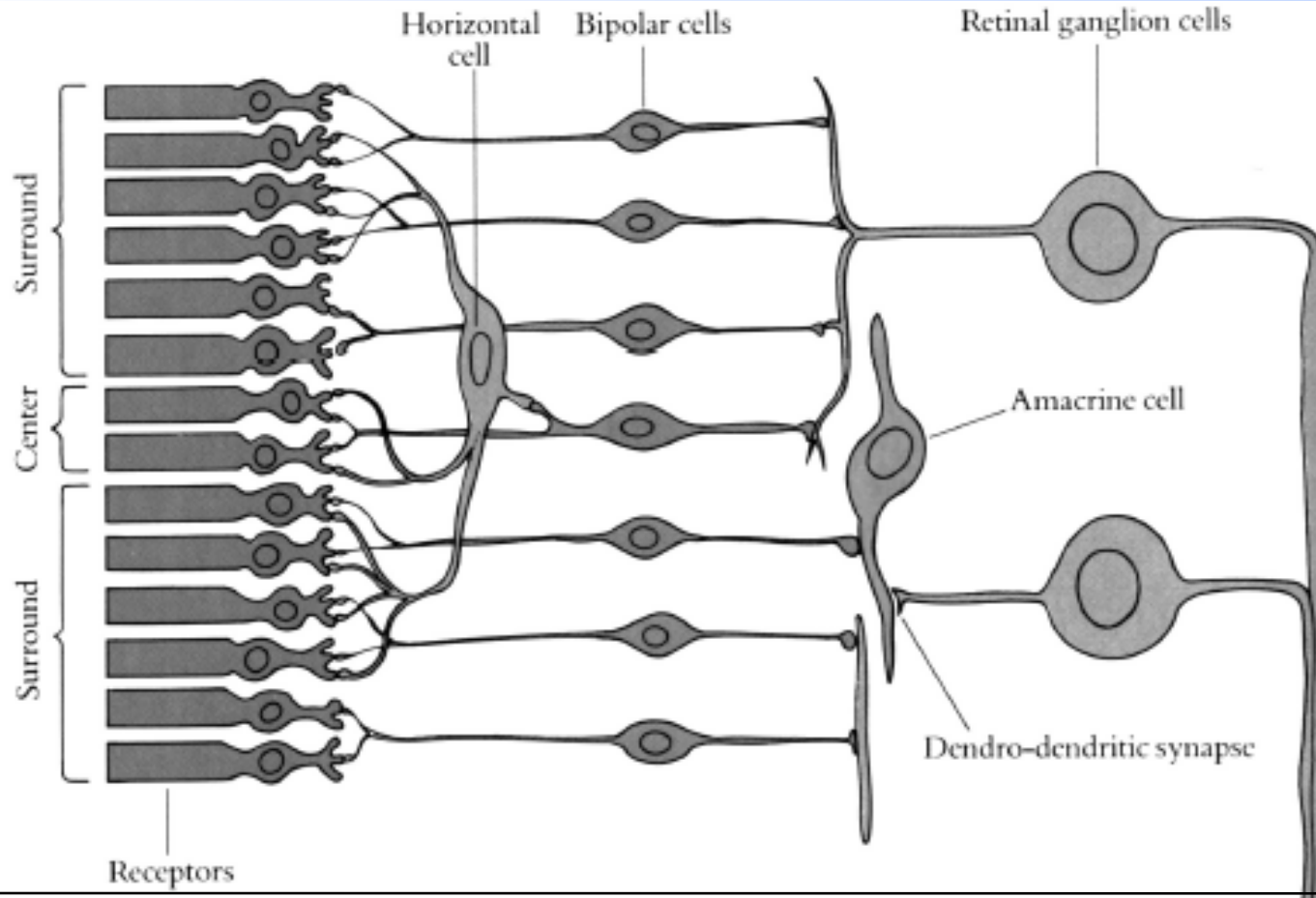


Center-surround receptive fields



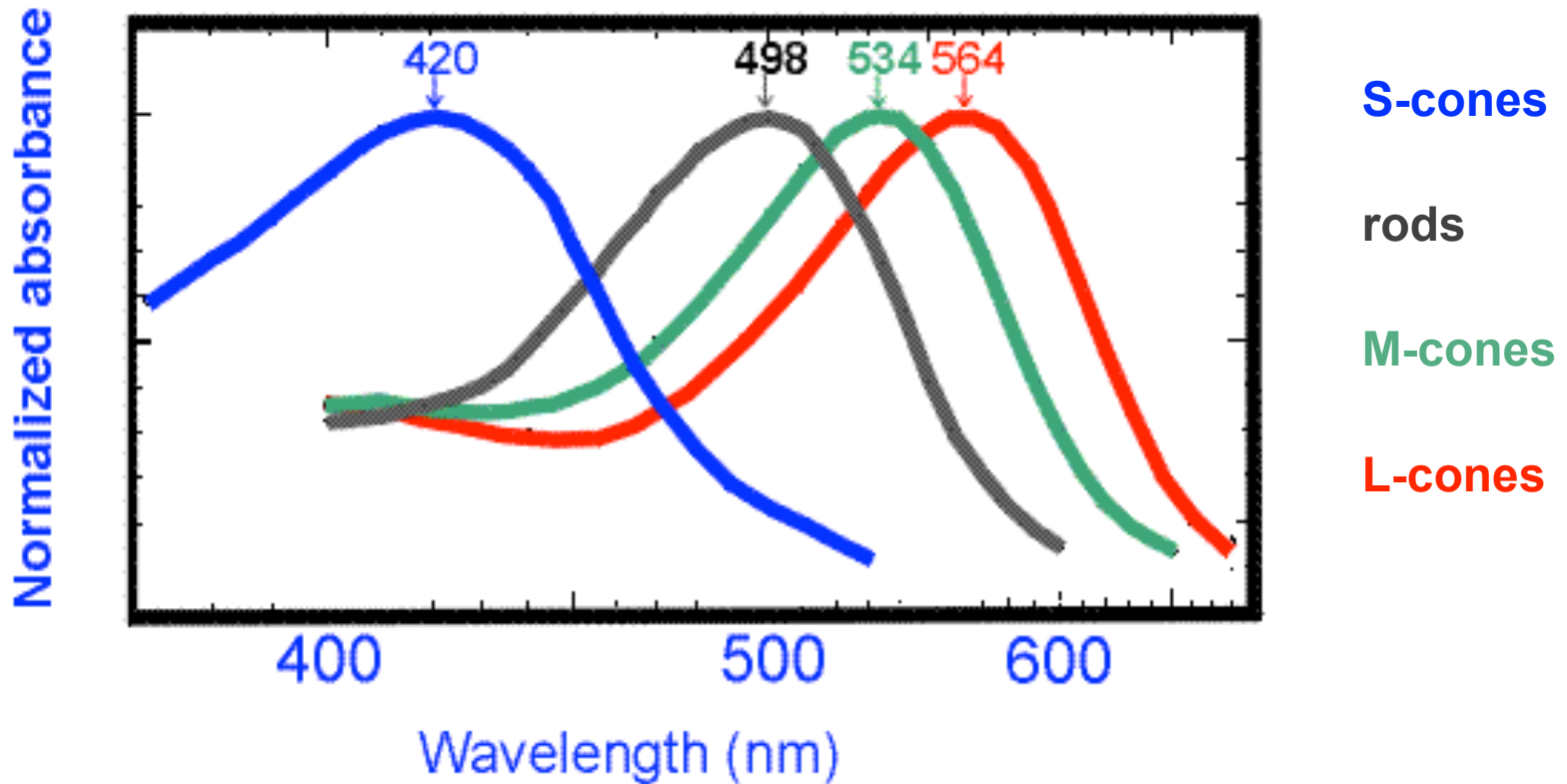


Generation of center-surround fields





Absorbance spectra of photo pigments



After Bowmaker & Dartnall, 1980



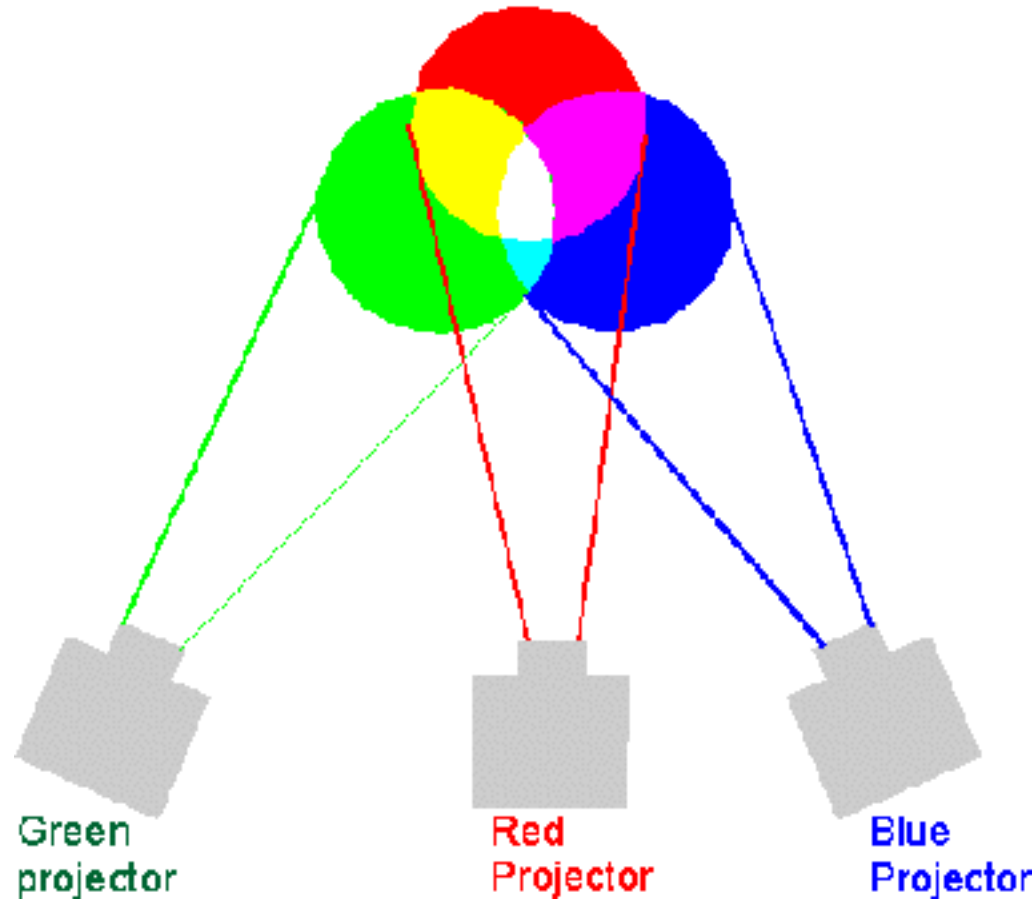
Colour vision theories

The **trichromatic** theory operates at the receptor level

The **opponent processes** theory applies to the subsequent neural level of colour vision processing

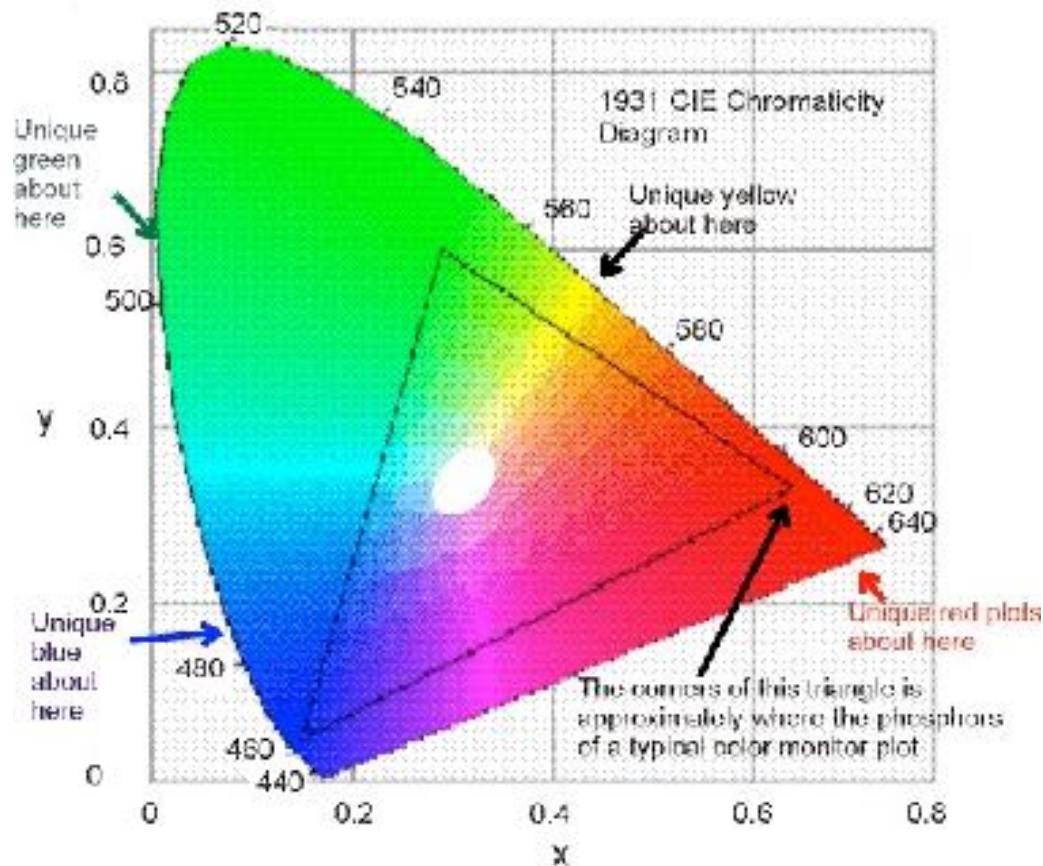


Additive colour mixing



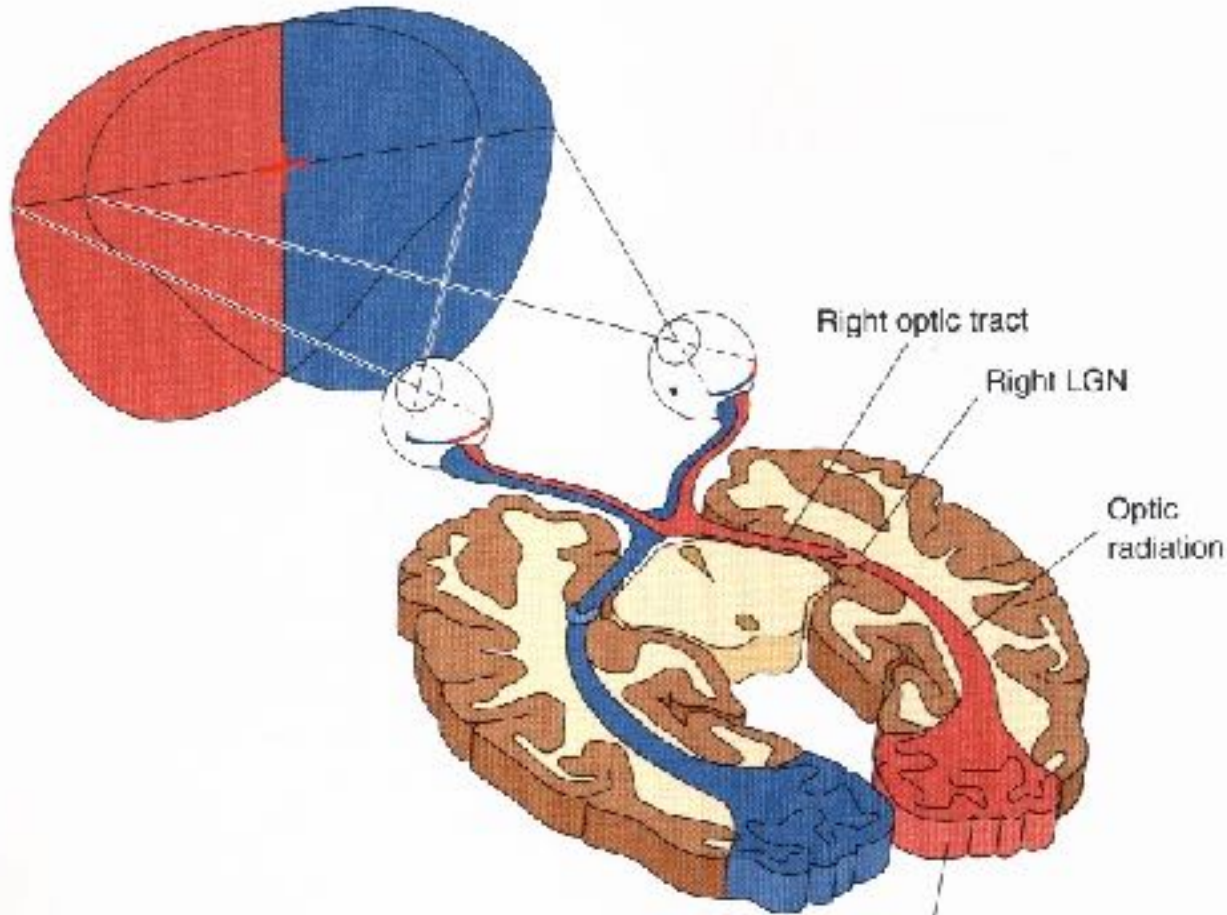


The CIE colour diagram



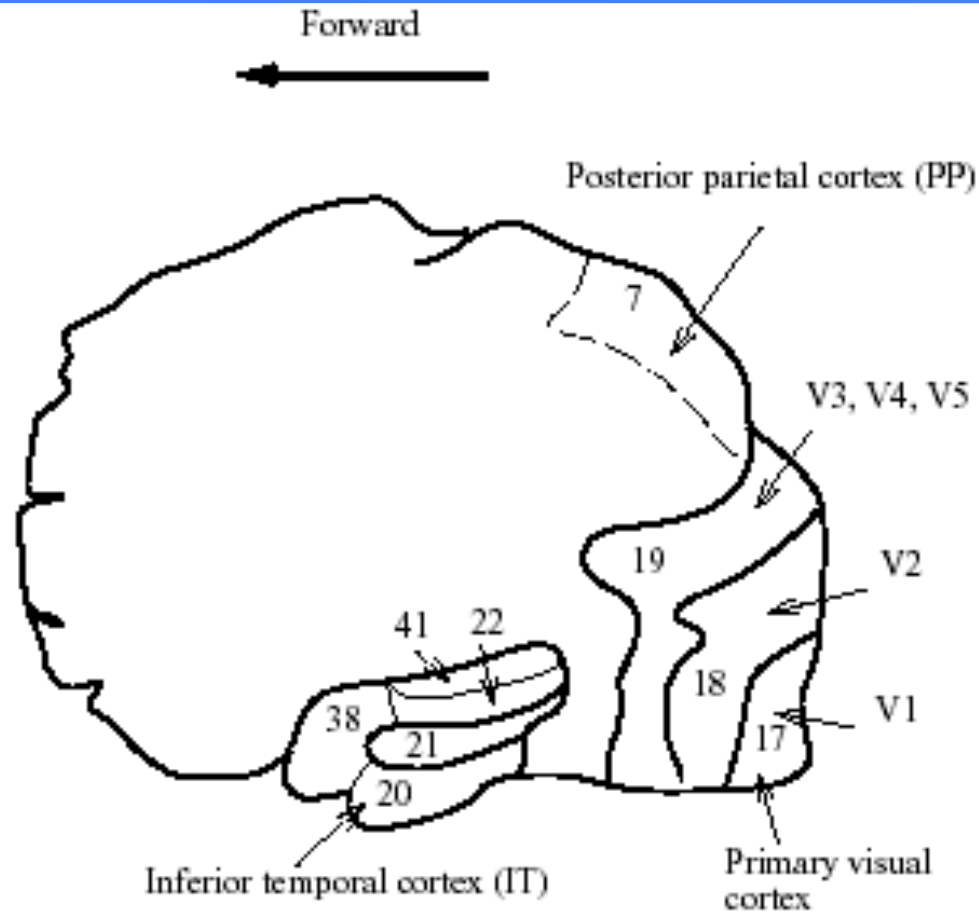


The visual pathway





Cortical maps





1981 Nobel prize in Medicine



David Hubel, Harvard

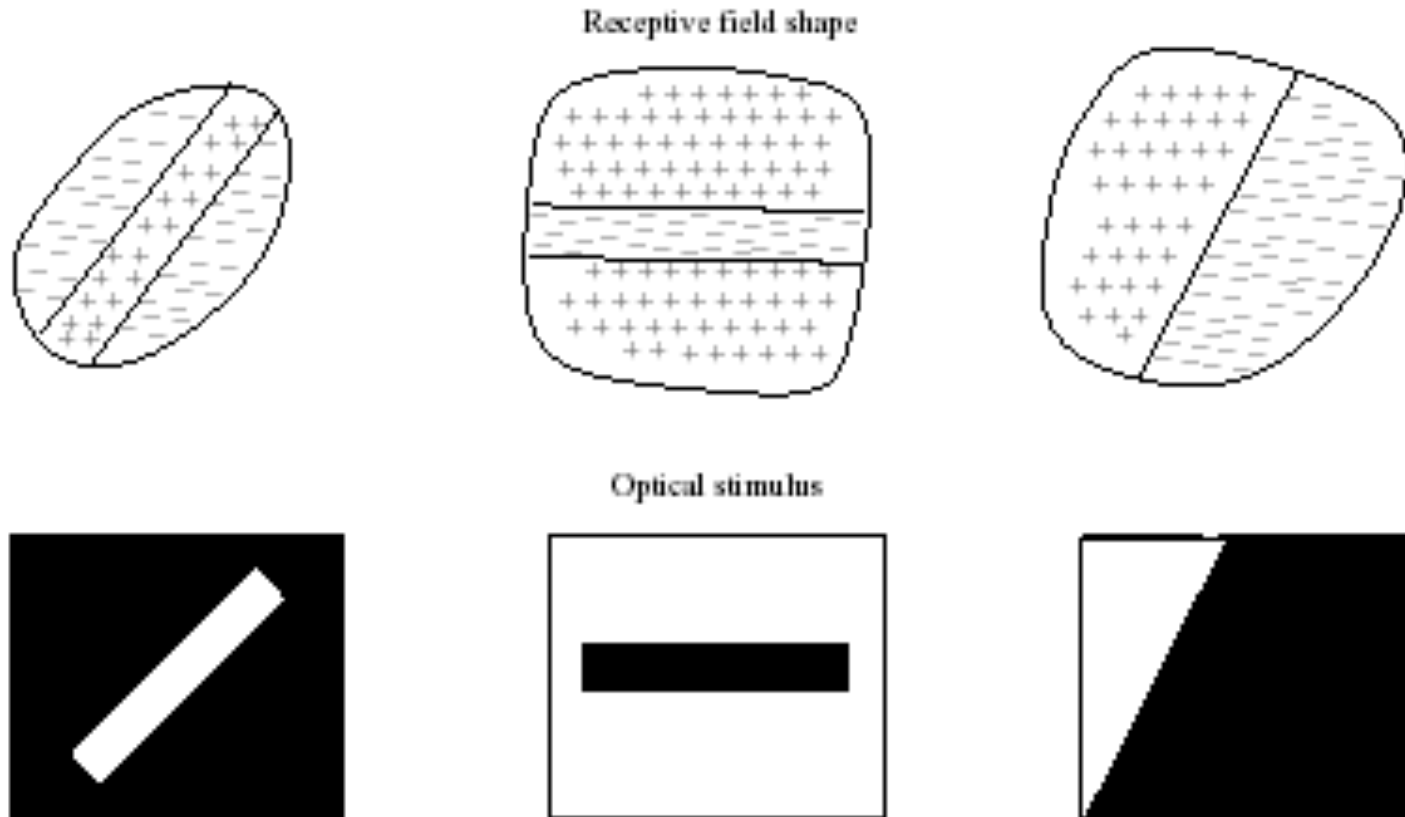


Torsten Wiesel, Harvard (initially KI)

- Microelectrodes in primary visual cortex of anesthetized cats
- What visual patterns are a particular cell sensitive to?



Receptive fields of simple cells





Preference of orientation and direction

Preferred orientation and direction



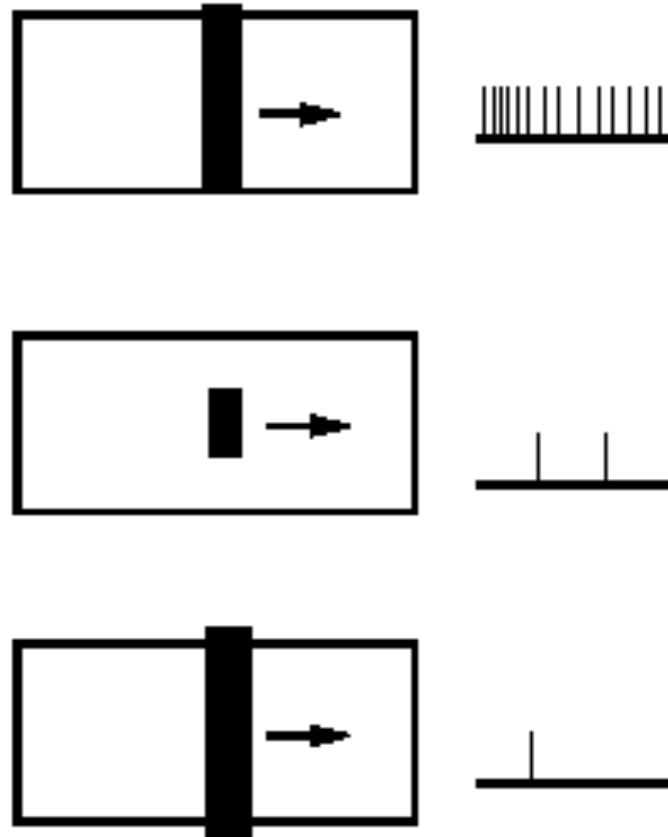
Preferred orientation and non-preferred direction





Length detector

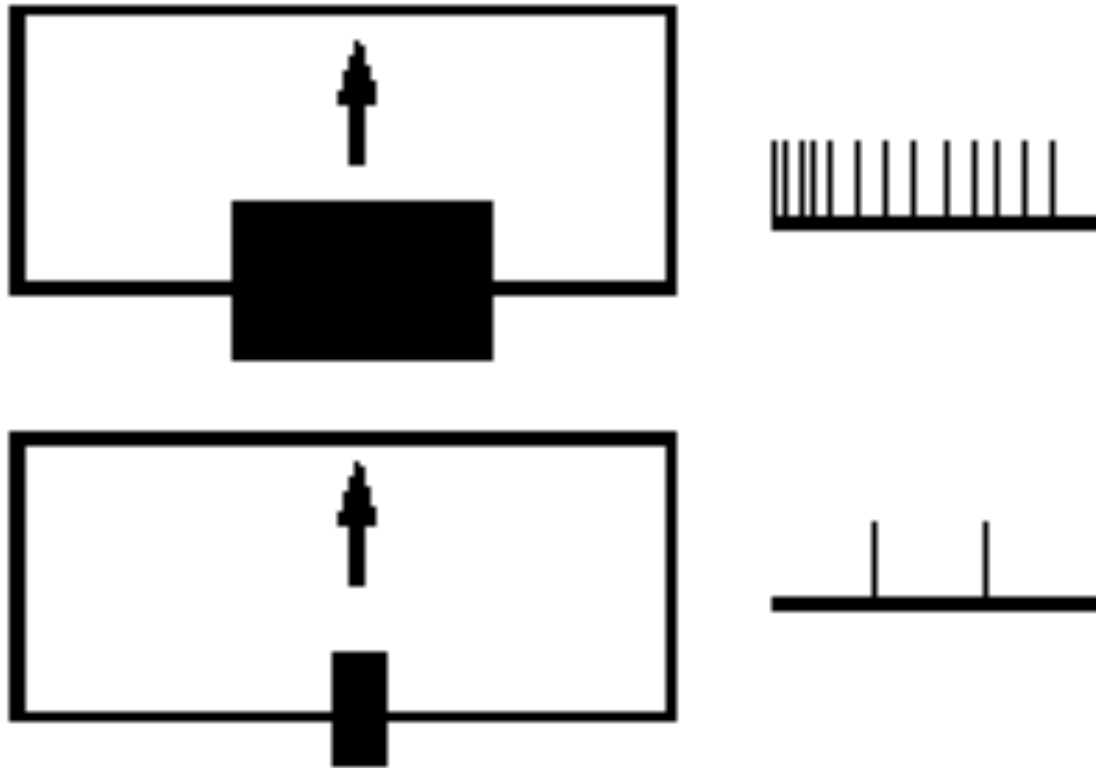
Length detector





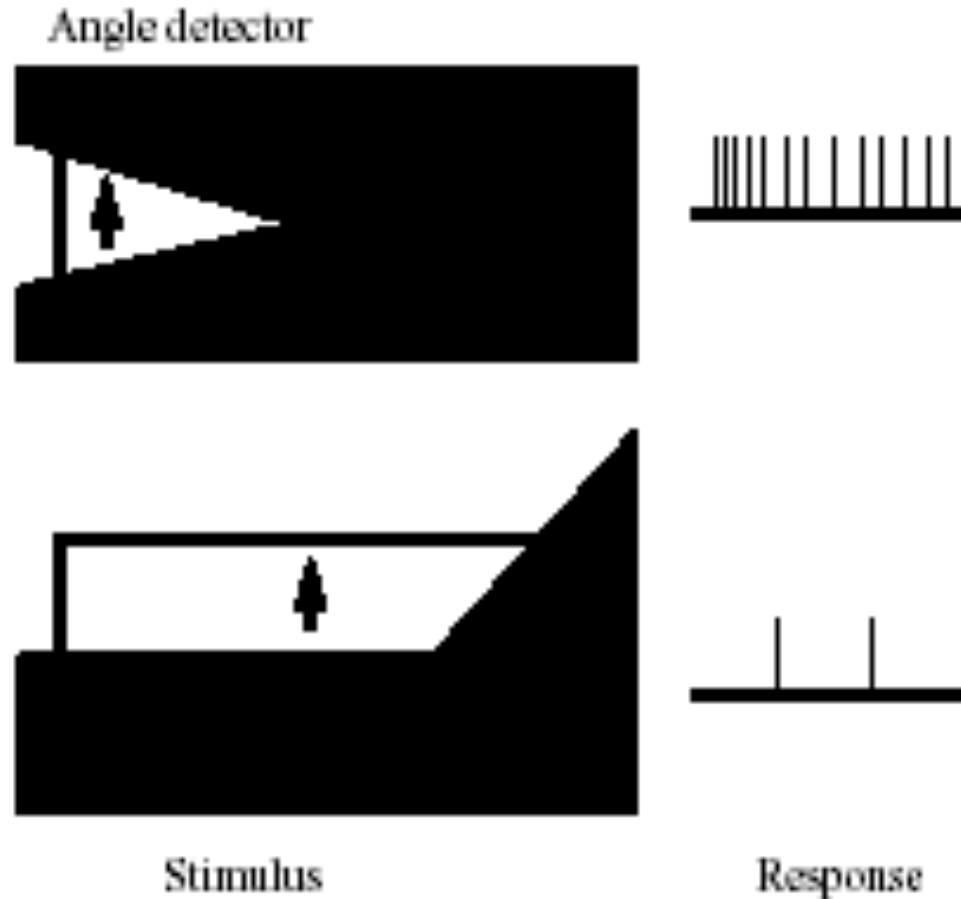
Width detector

Width detector





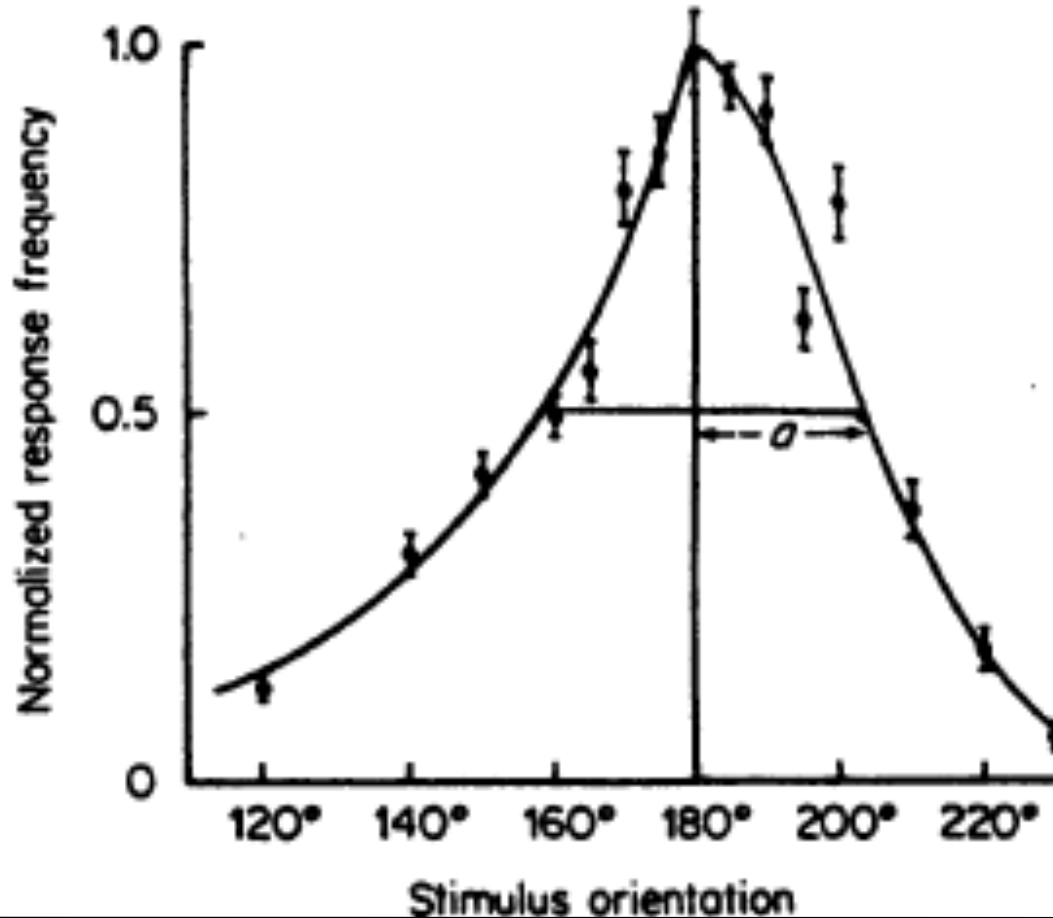
Angle detector





Orientation tuning

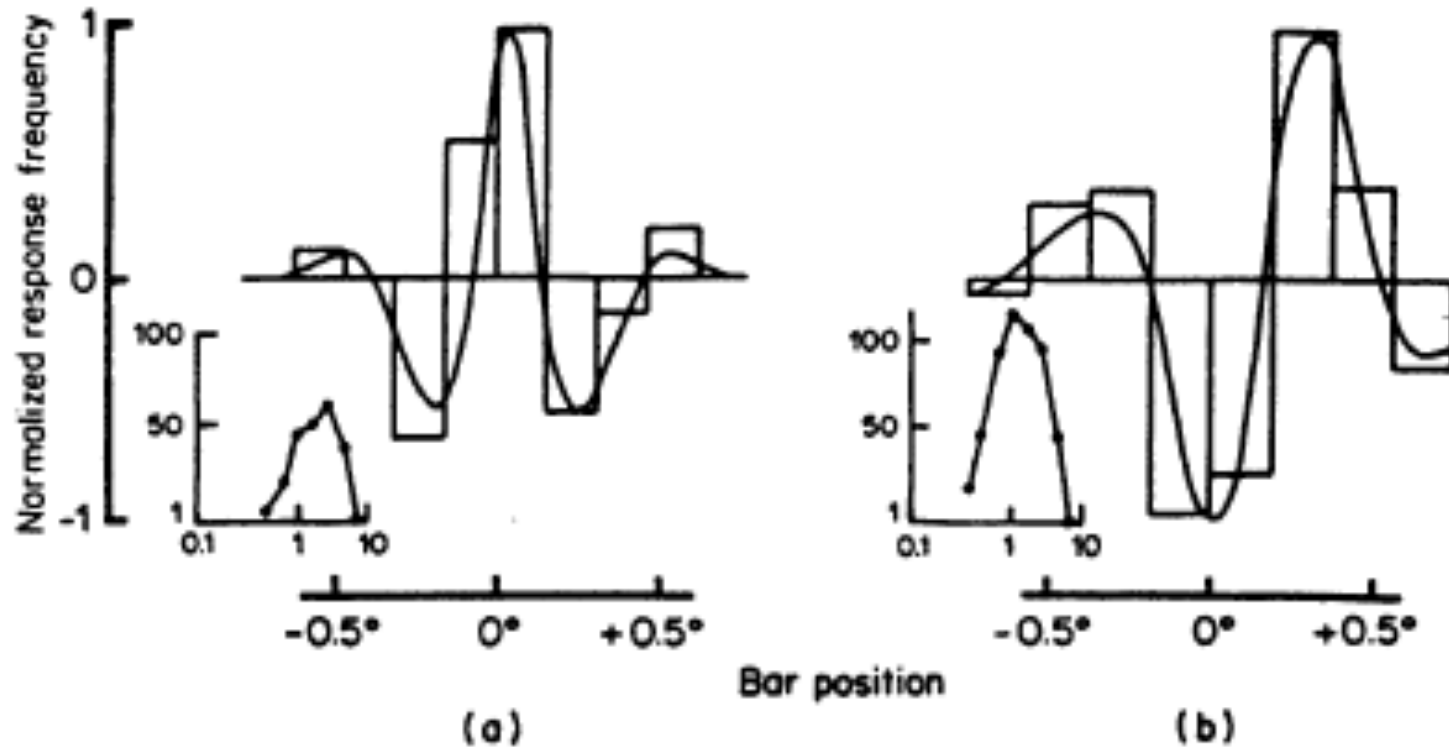
Simple cell of cat





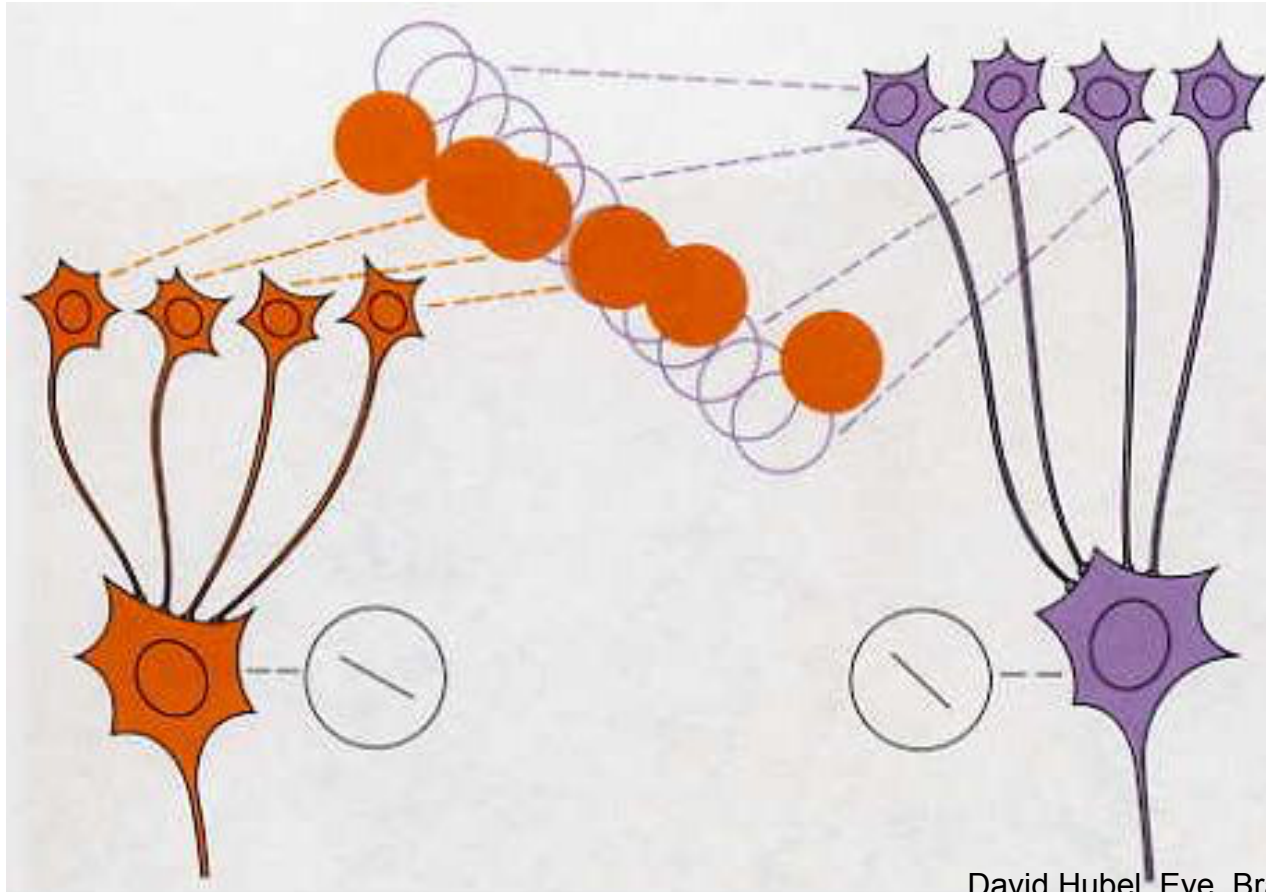
Sensitivity profiles of simple cells

a) Bisymmetrical b) Antisymmetrical





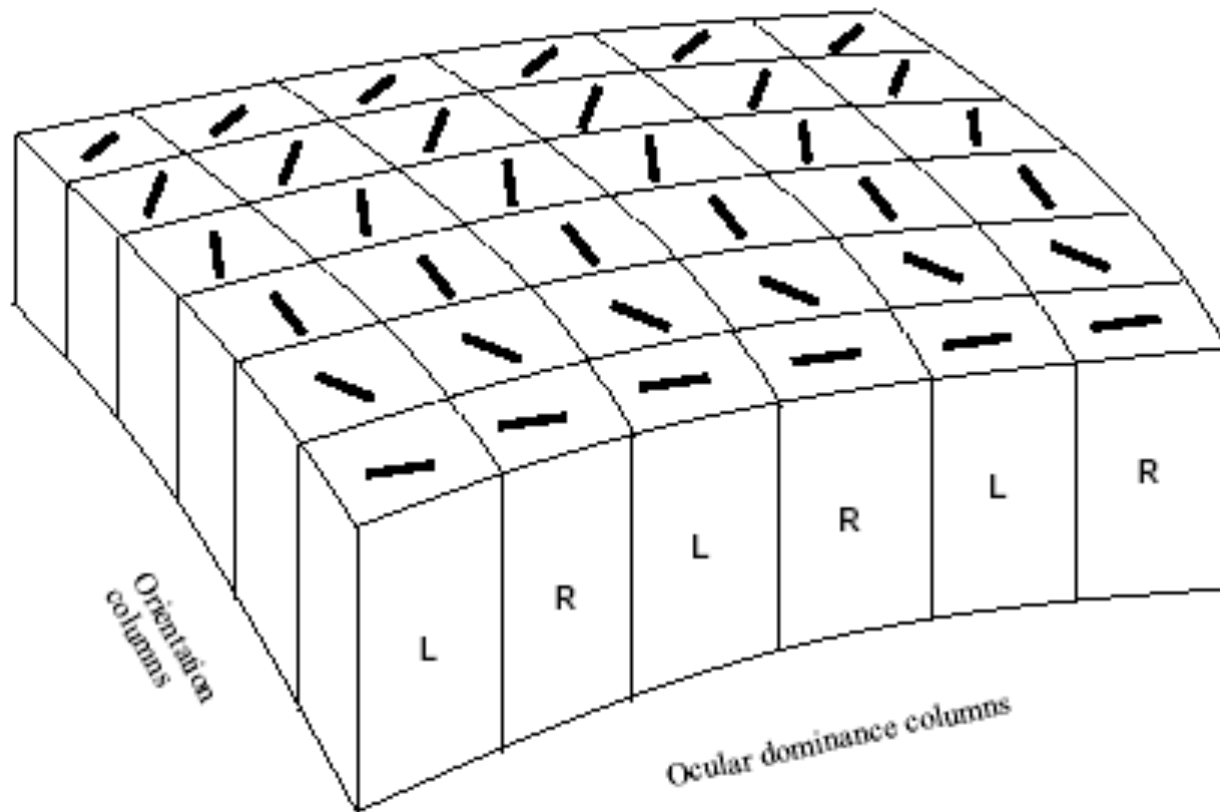
Implementation of simple cell receptive fields



David Hubel, Eye, Brain and Vision

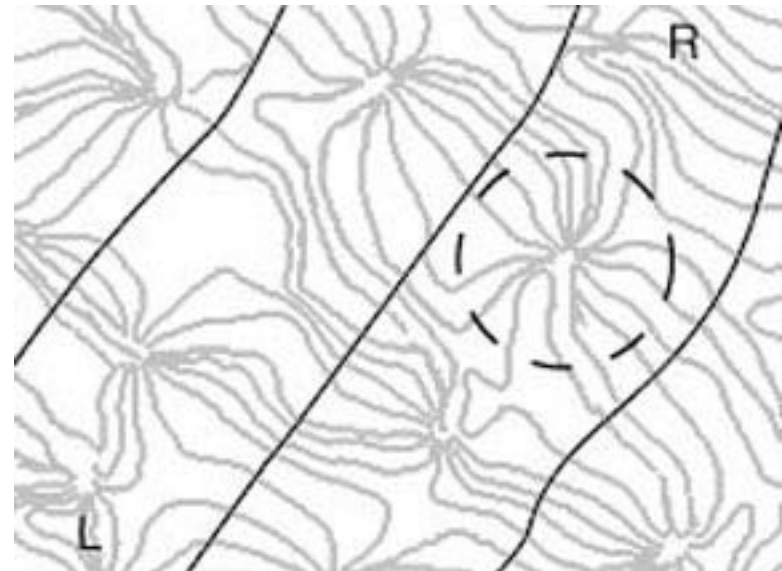
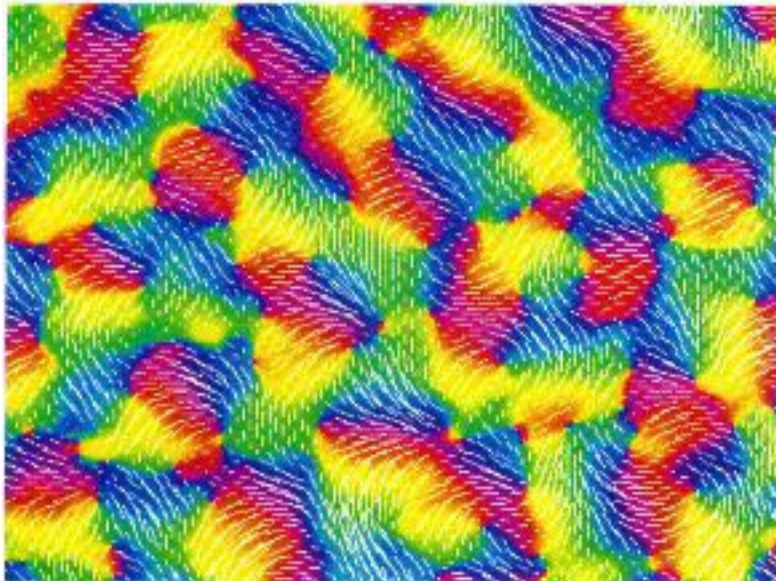


Orientation and ocular dominance columns



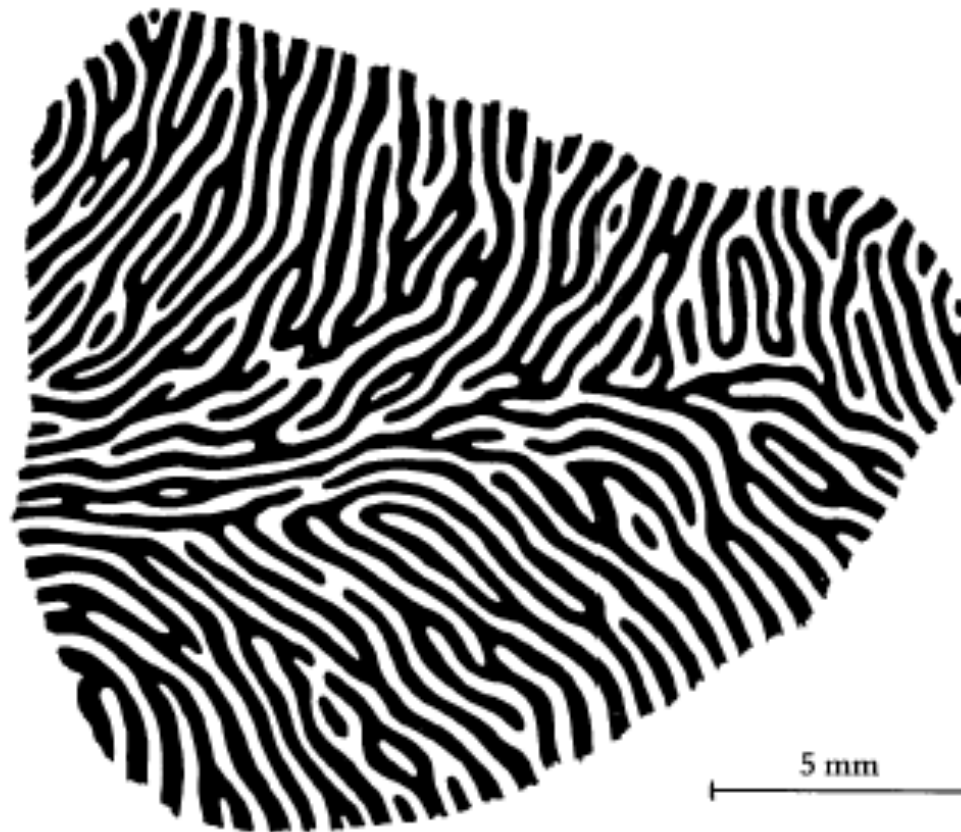


Orientation dominance



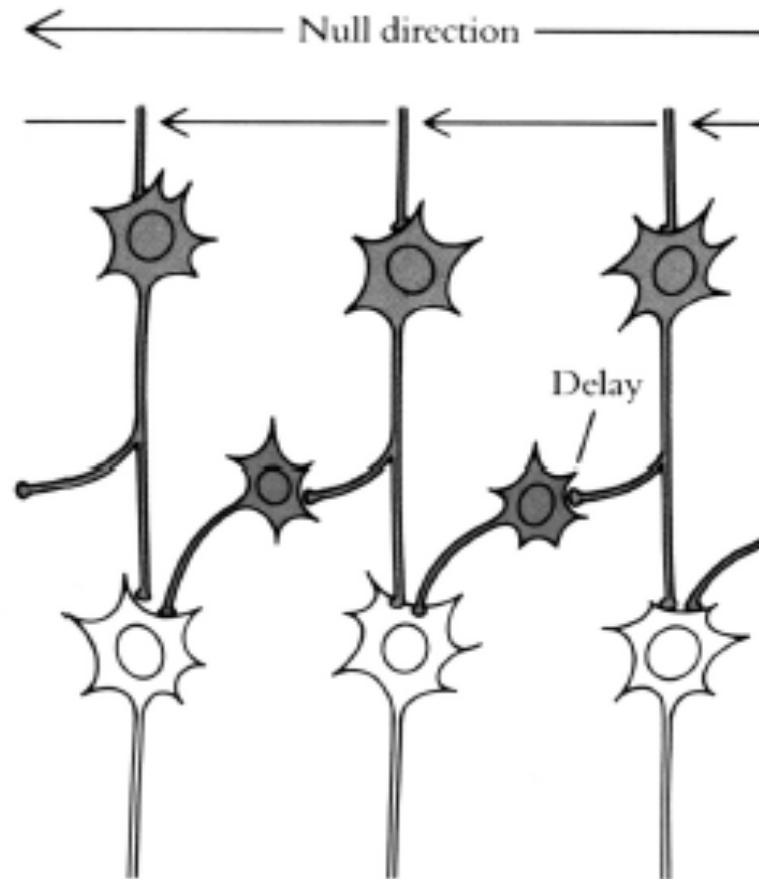


Ocular dominance map



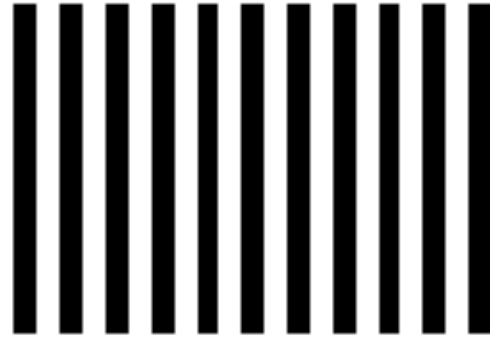
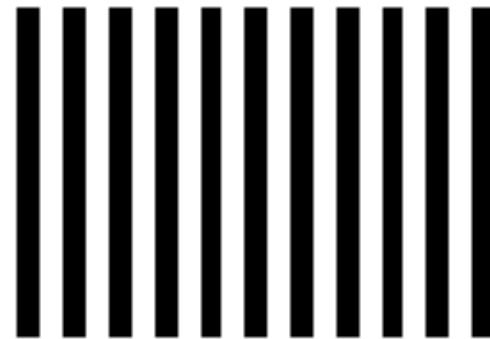
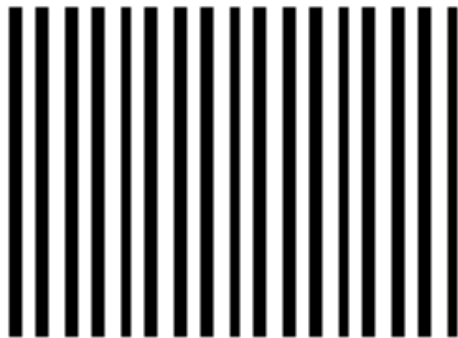


Implementation of direction-sensitive cell





Spatial frequency adaptation

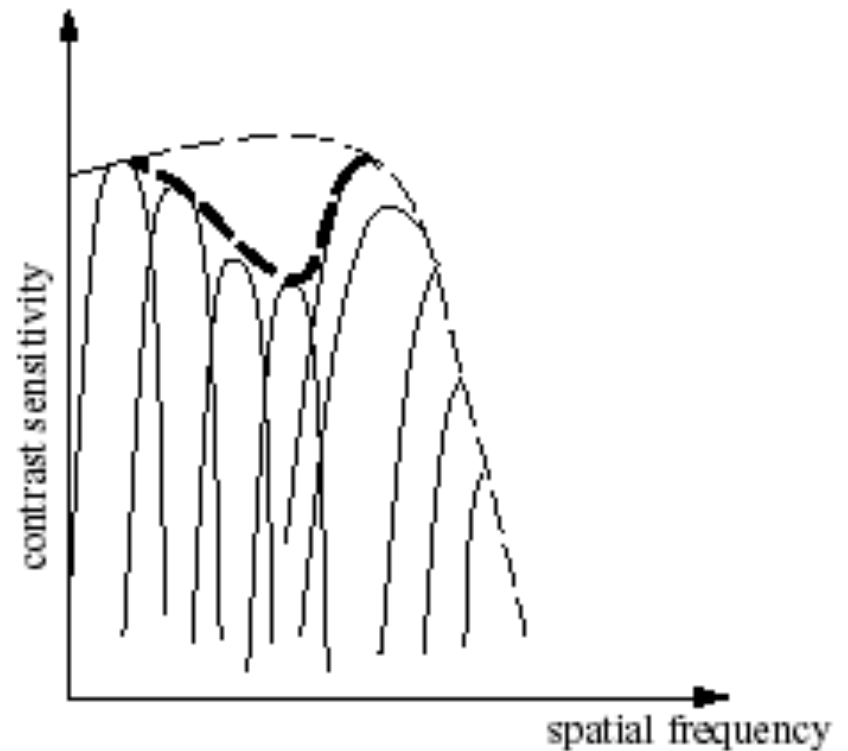
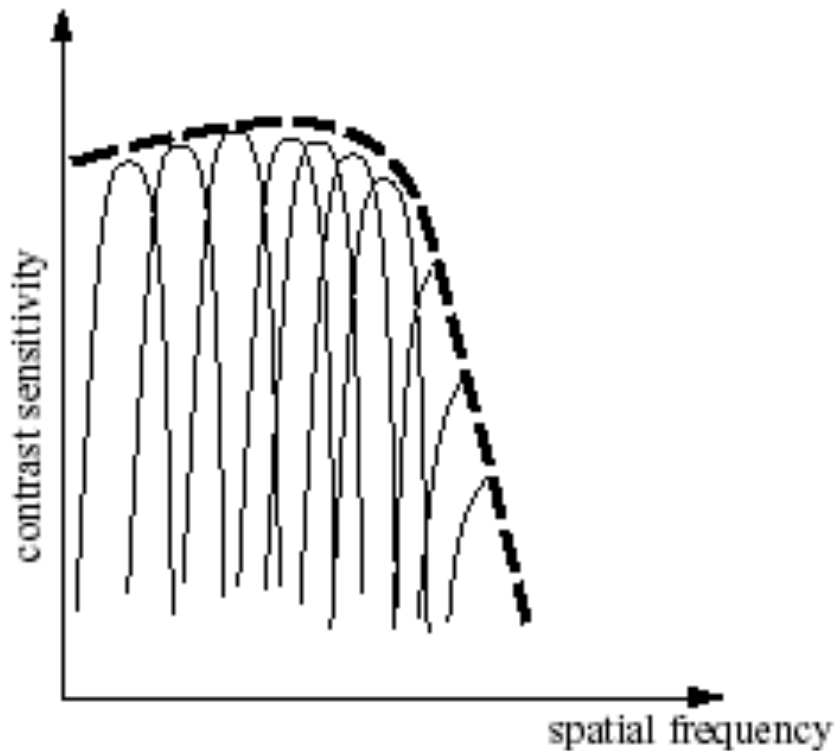


Adapted from Blakemore & Sutton, 1969



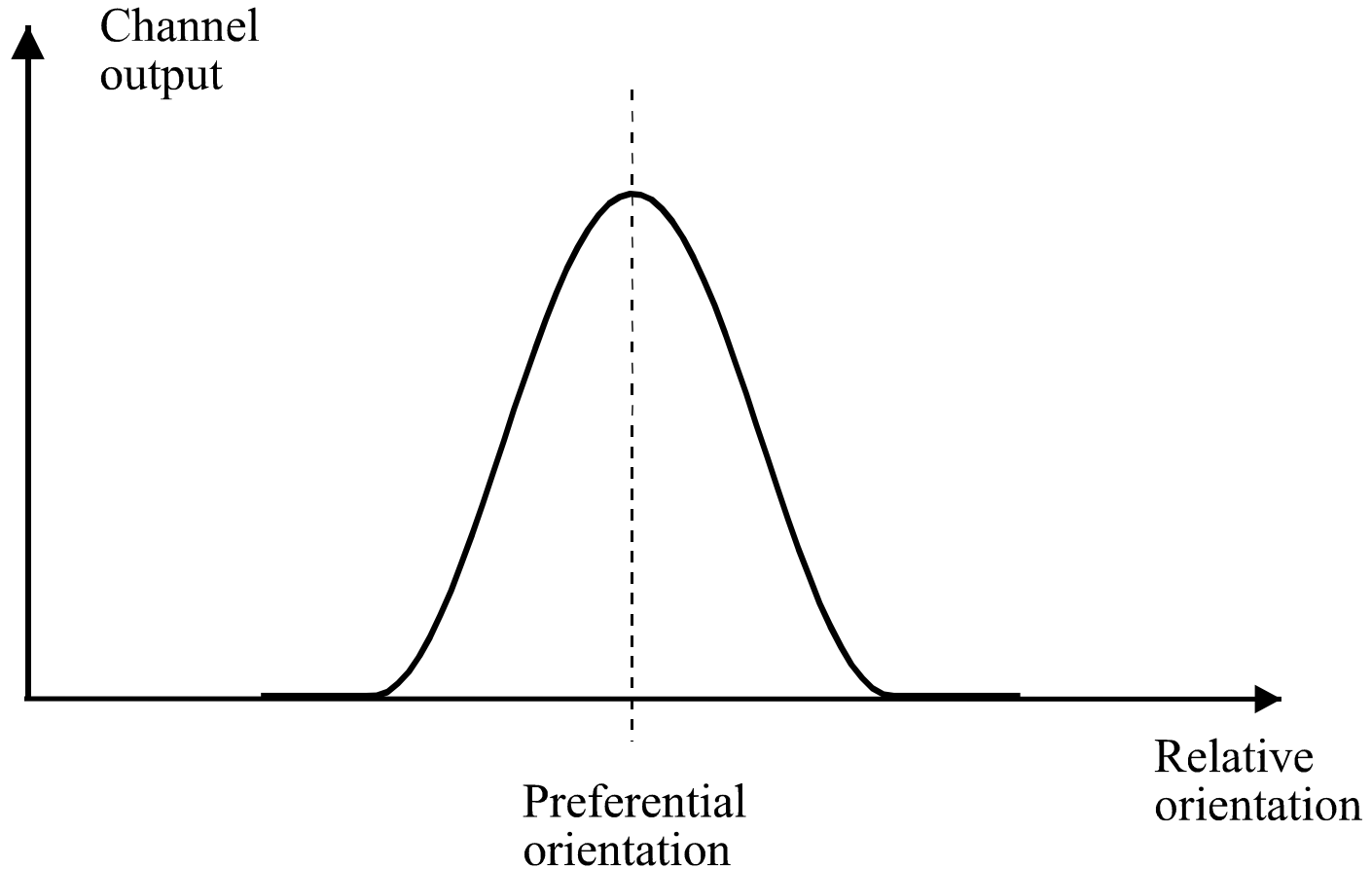
Build-up from separate channels

Effect on sensitivity of channels



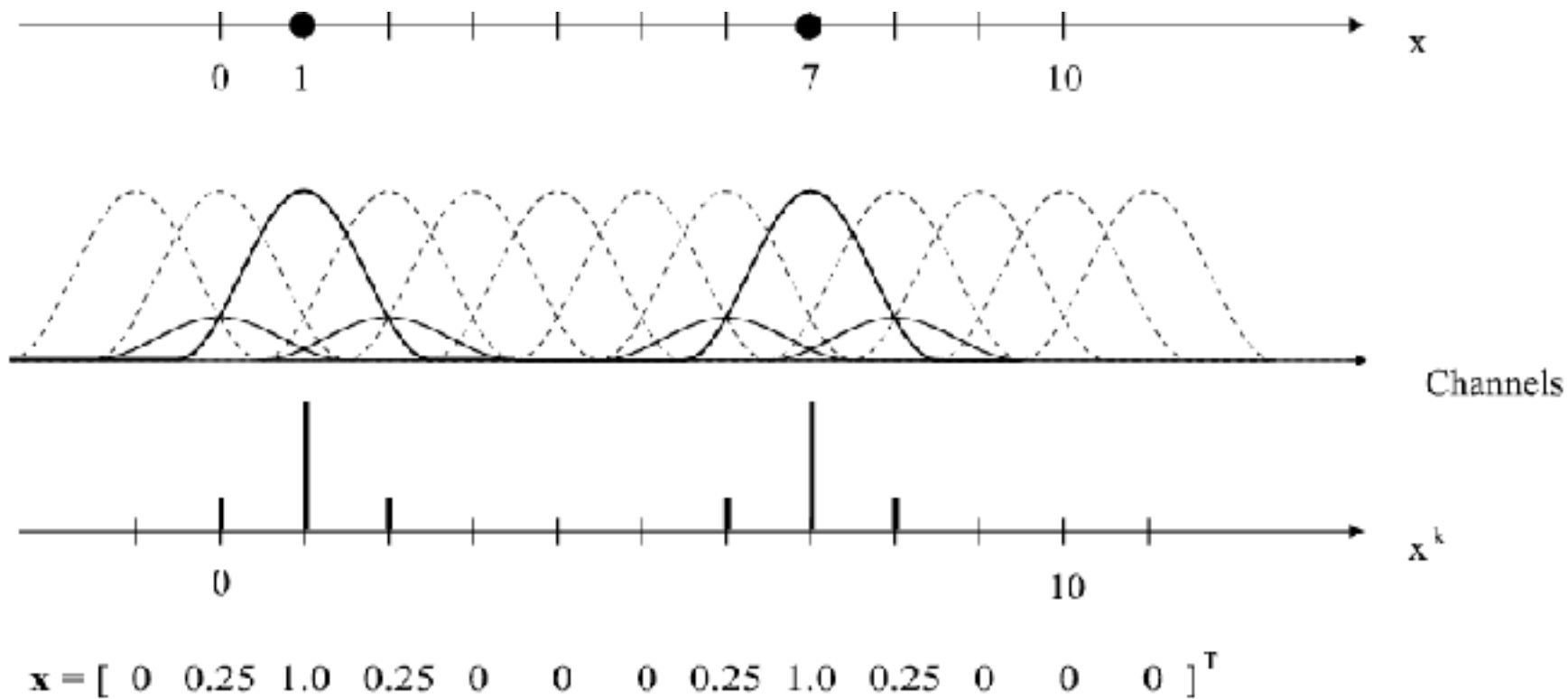


Channel representation





Channel Information Representation



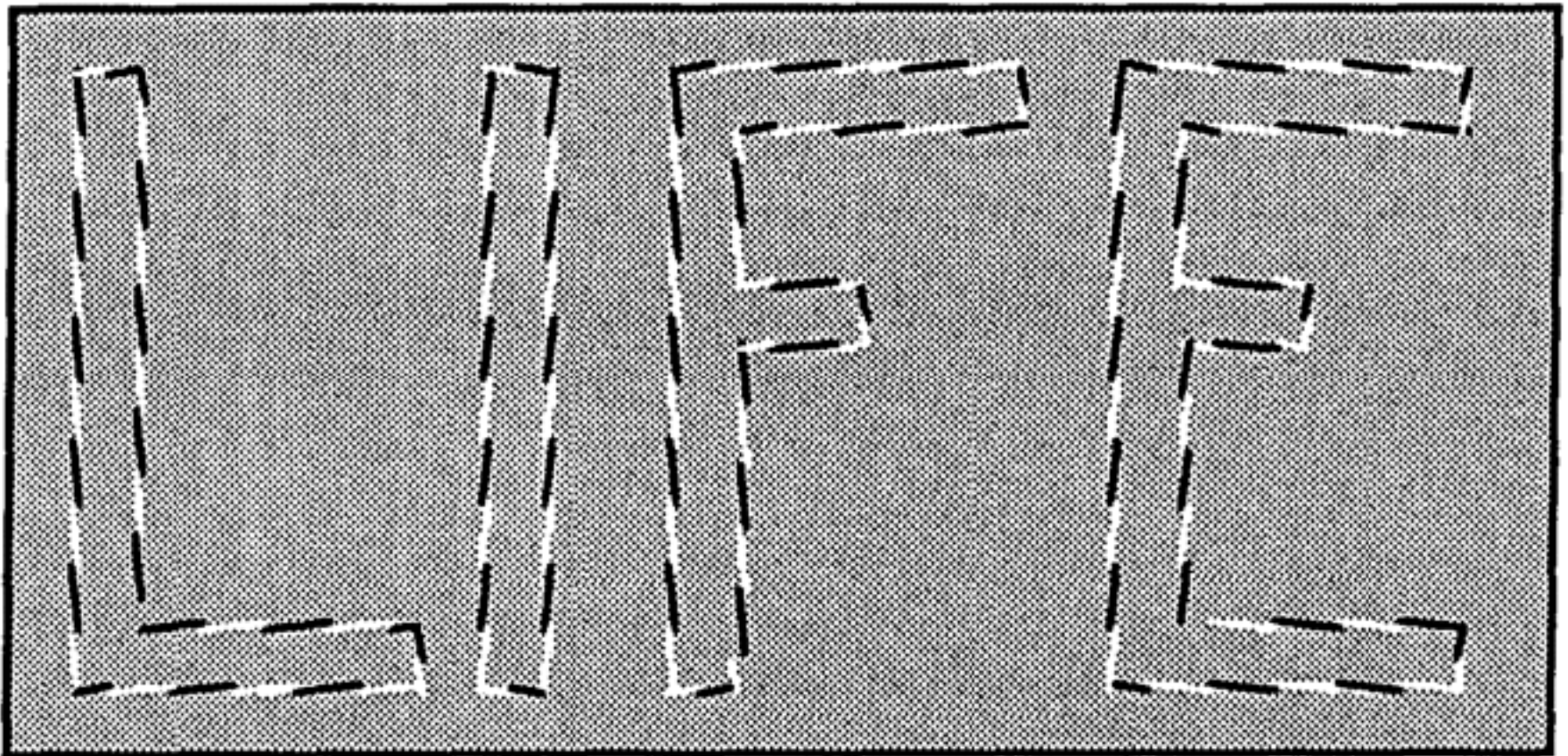


Advantages of channel representation

- Several values can be represented for a variable, allowing support to alternative hypotheses
- Locality allows a fast optimization in learning
- Locality allows implementation of non-linear models using linear mappings
- Allows representation of confidence or certainty
- Monopolarity allows **zero** to represent **no information** (i.e. "I don't know")
- Sparse representation



Local versus global properties





Conflicting interpretations





Parallel interpretation



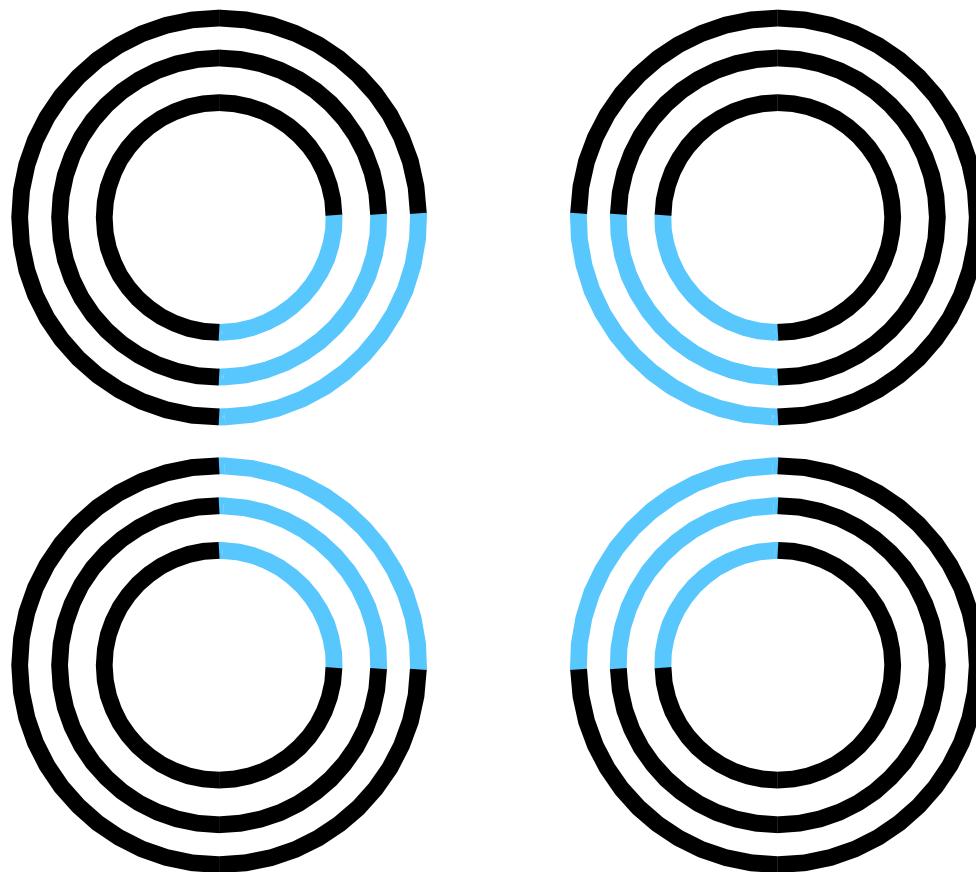


Sequential interpretation



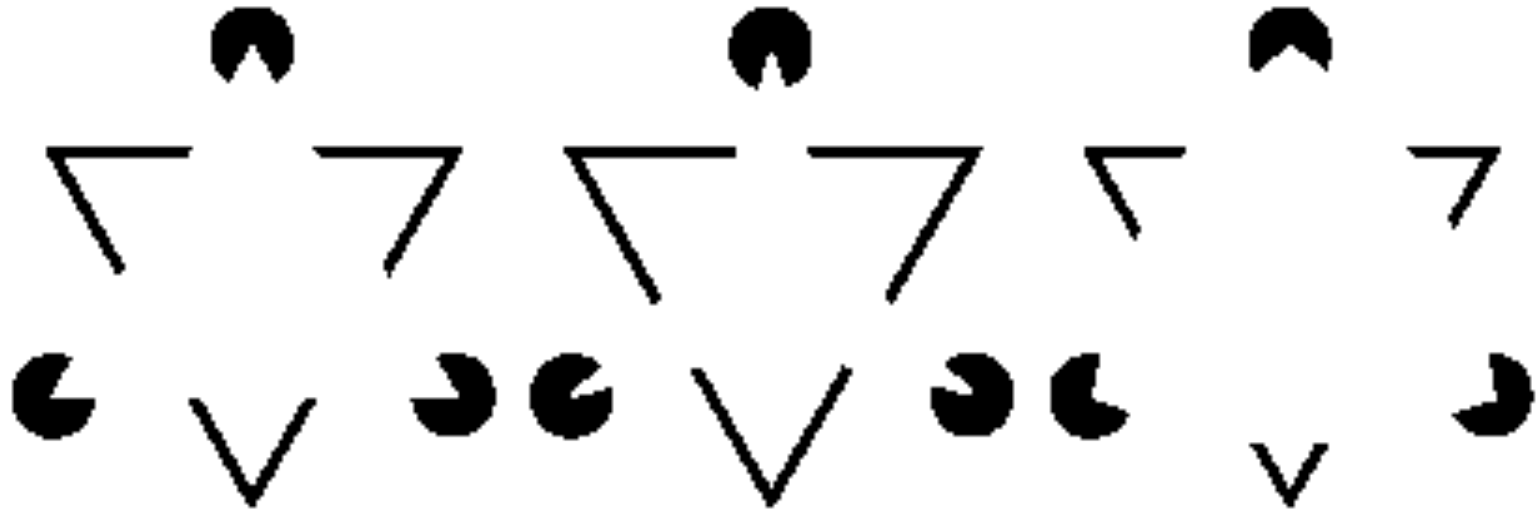


Extrapolations forming illusions



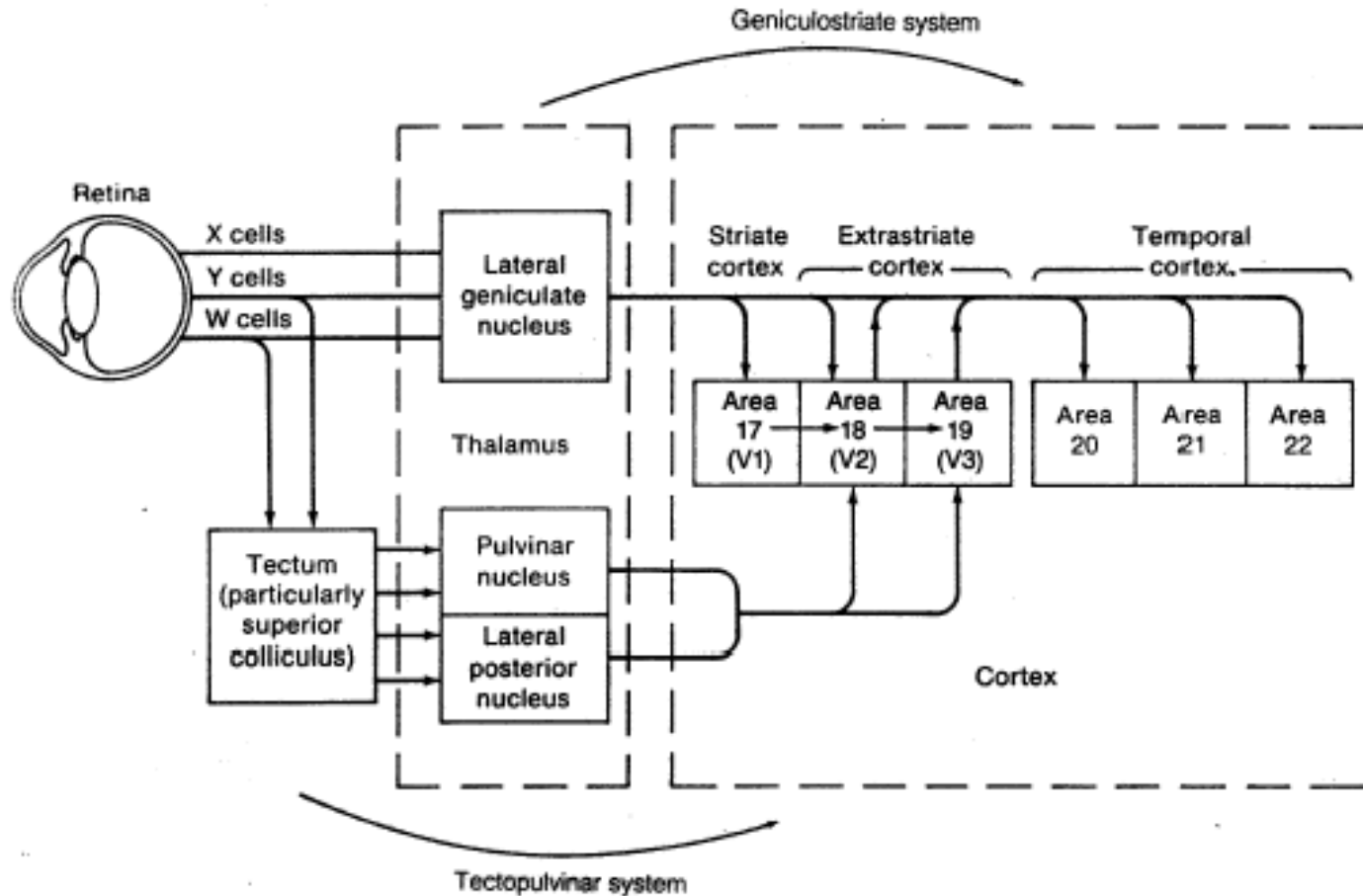


The Kanitza triangle



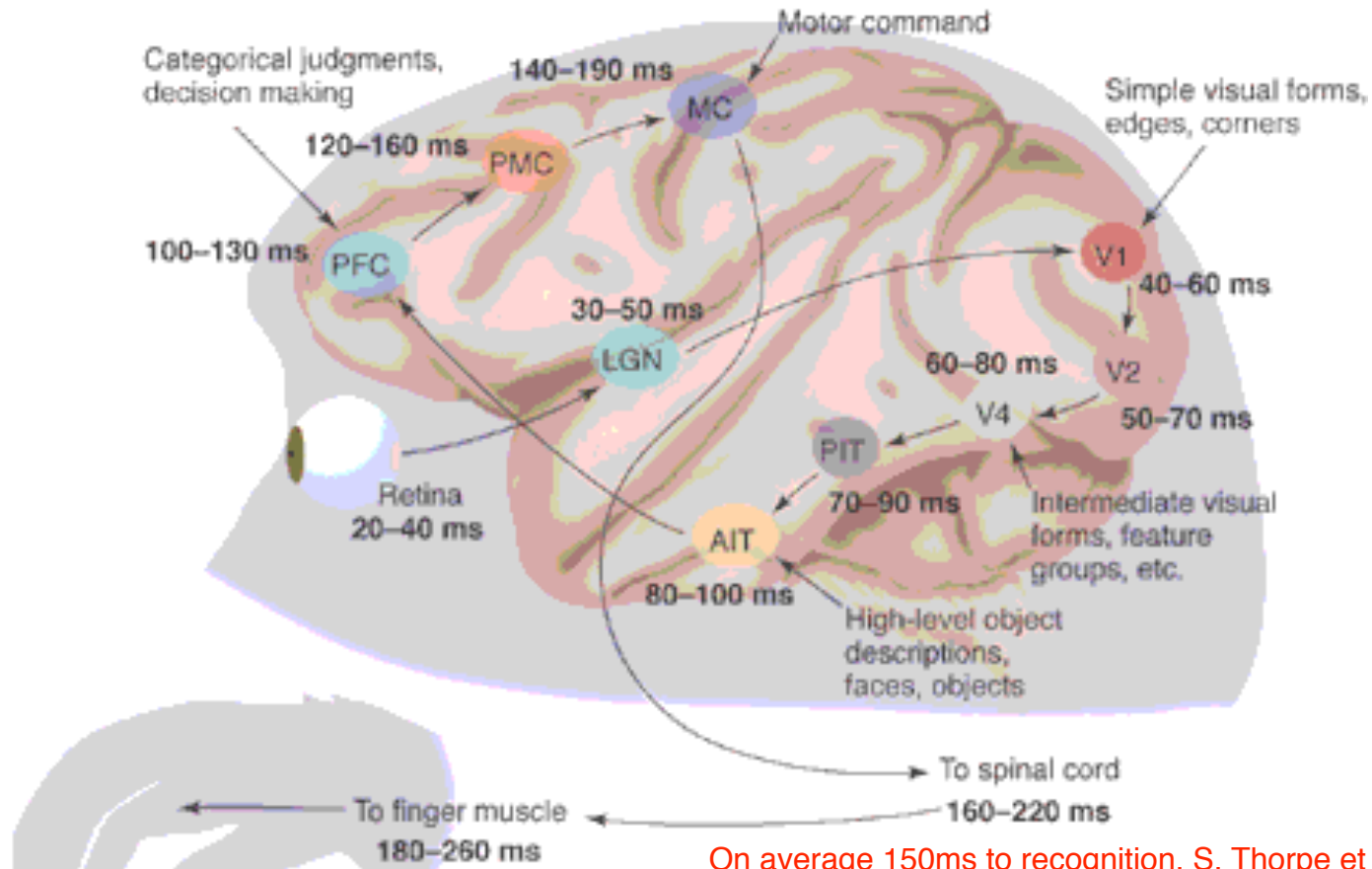


Part of processing pathway



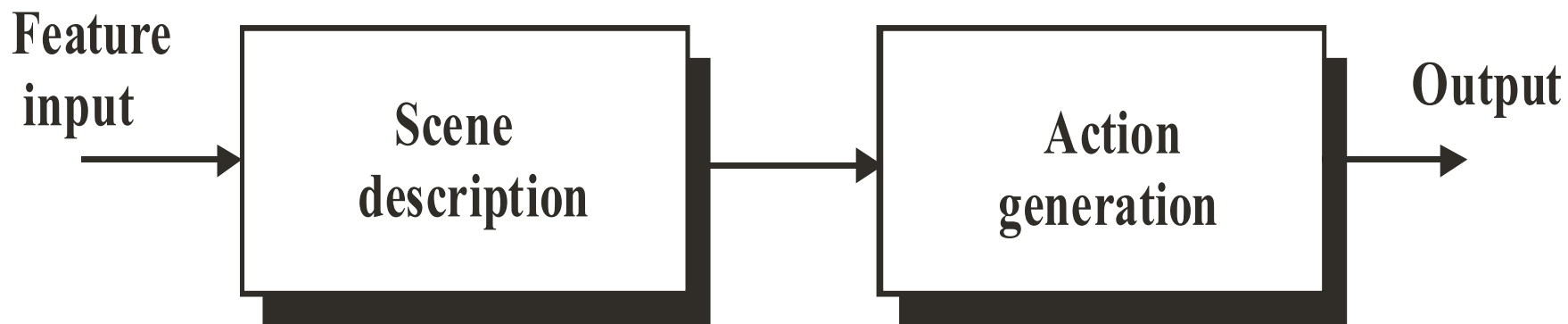


Computation times



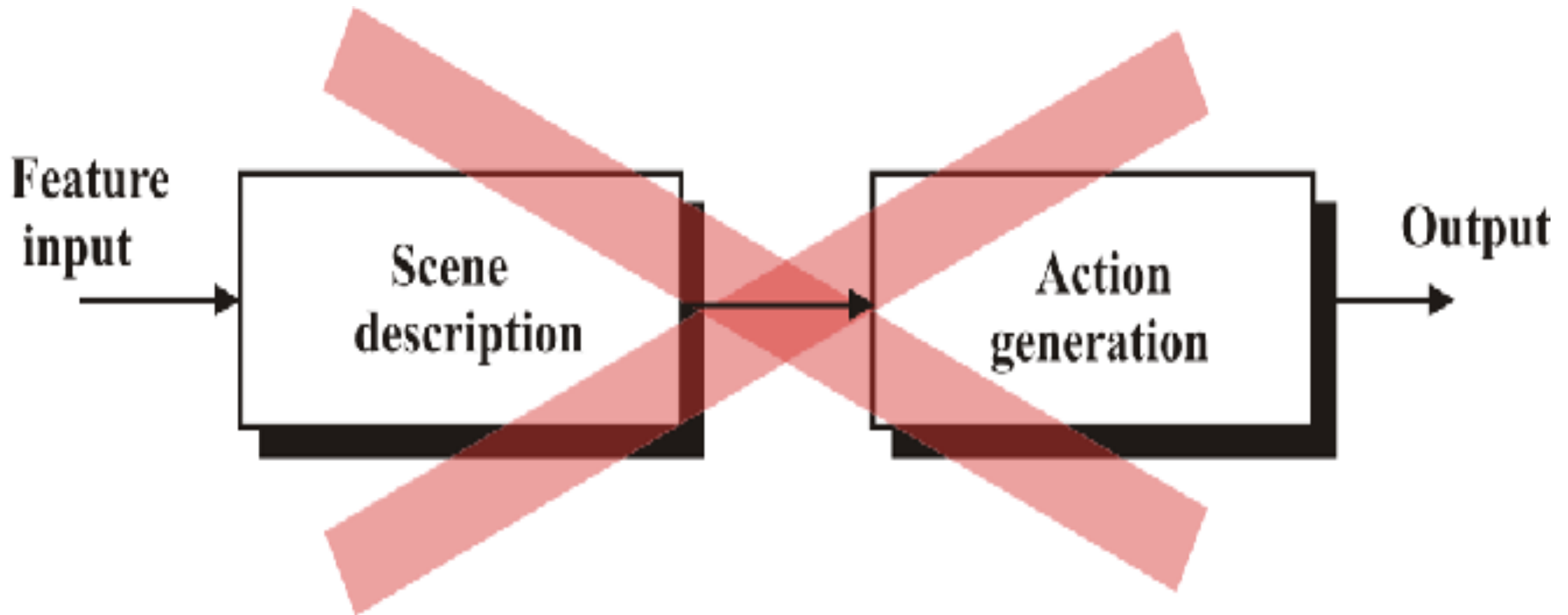


A conventional robotics structure





Not done in biological vision

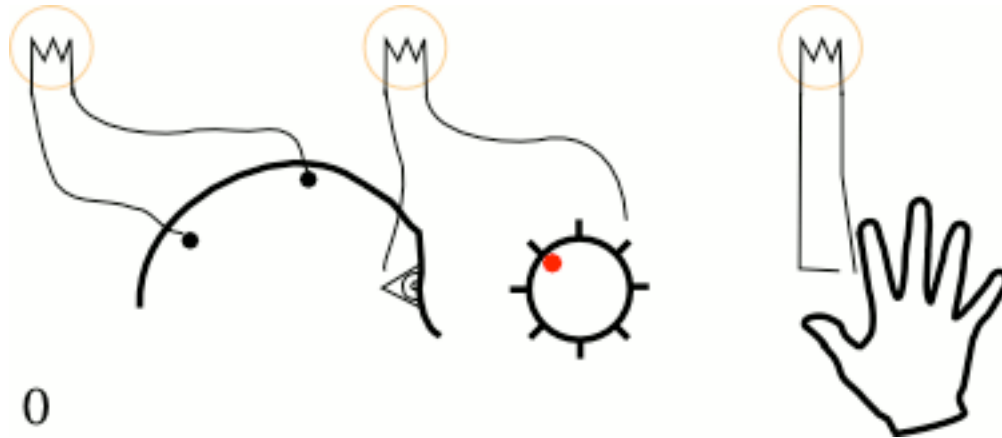




Consciousness - an afterthought

Experiments by Benjamin Libet show that:

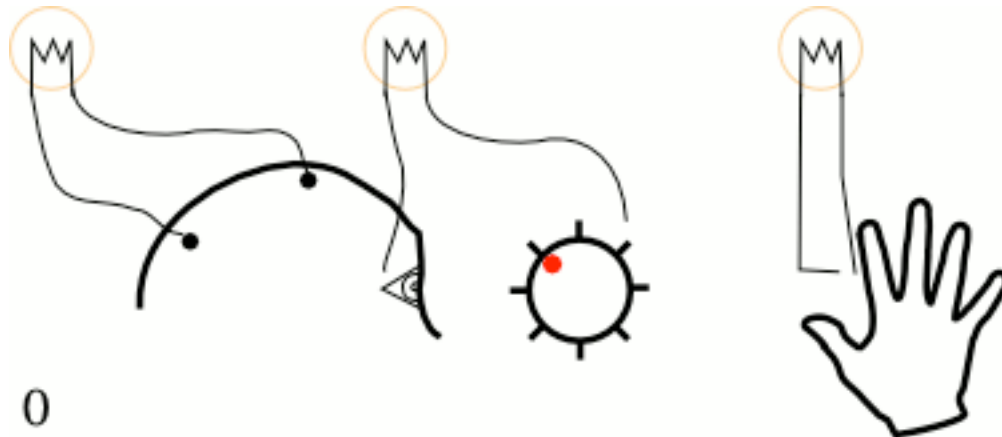
Action is initiated before it reaches consciousness





Consciousness - an afterthought

Synchronized EEG and rotating clock, subjects note position on timer when "they were first aware of the wish or urge to act"



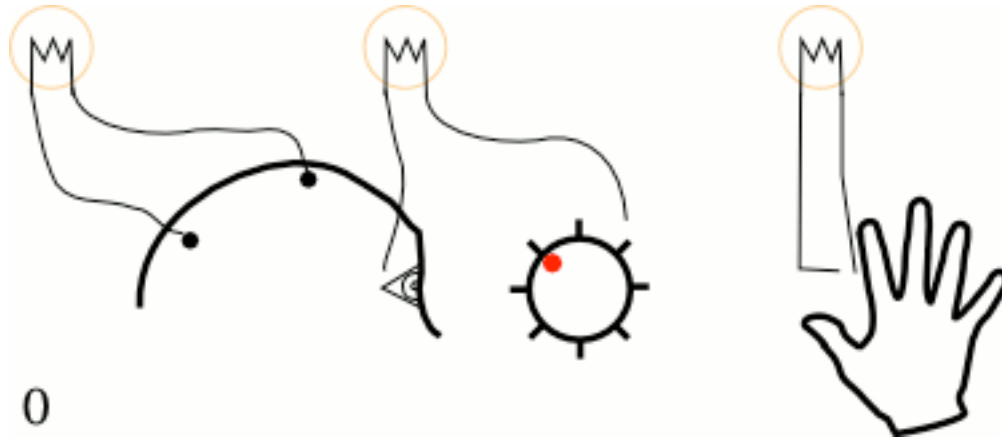


Consciousness - an afterthought

T-500ms: Readiness potential is measured by EEG

T-200ms: Observed time is registered by consciousness
by looking at synchronised clock

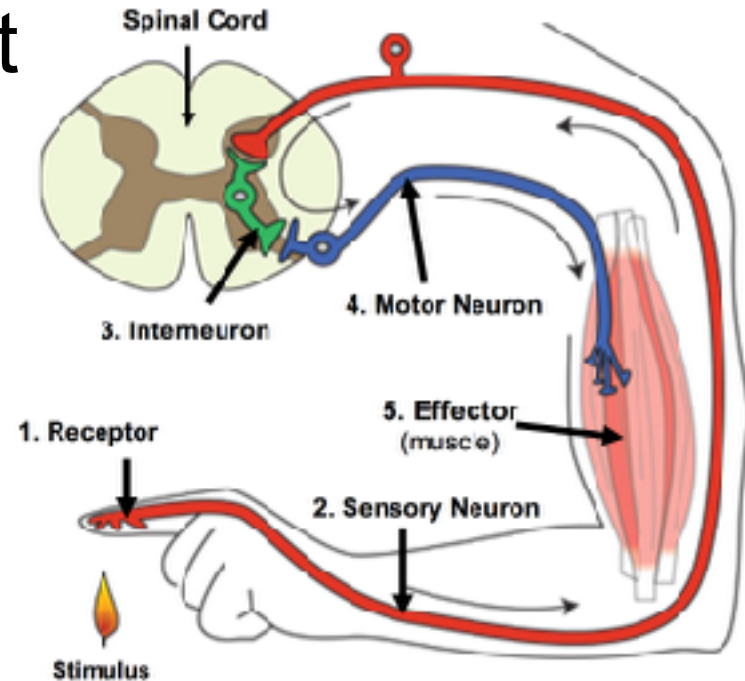
T: Action takes place





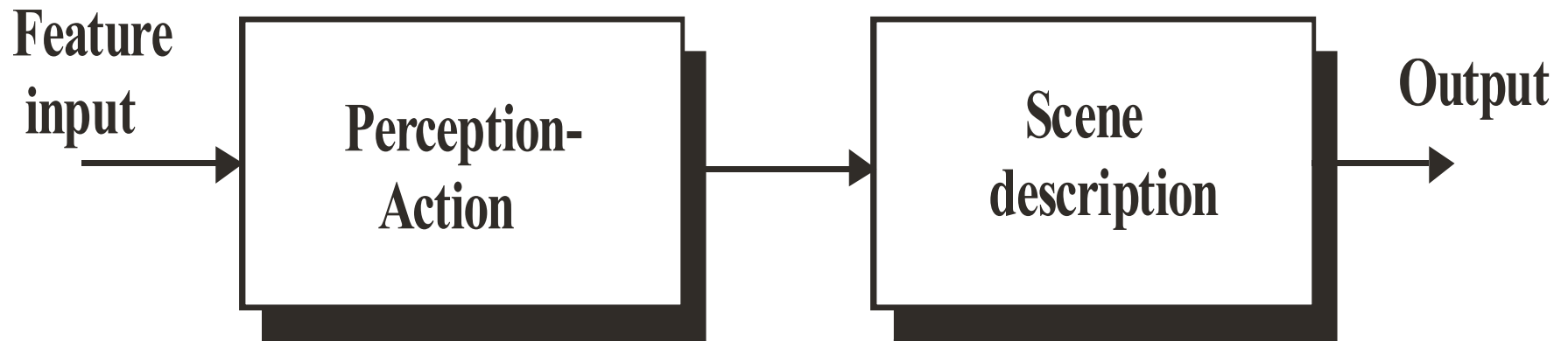
Other examples

1. It is well known that reflex actions are pre-conscious
2. You do not consciously plan all details of e.g. walking pattern



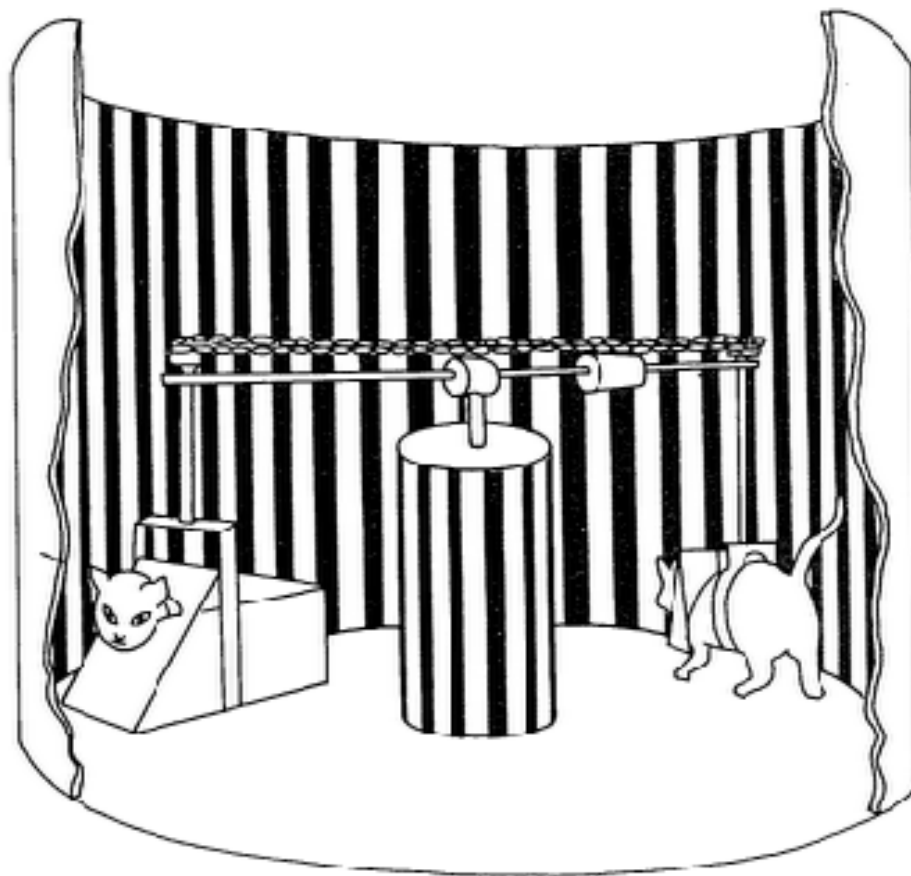


Order is the opposite!





Active versus passive exposure





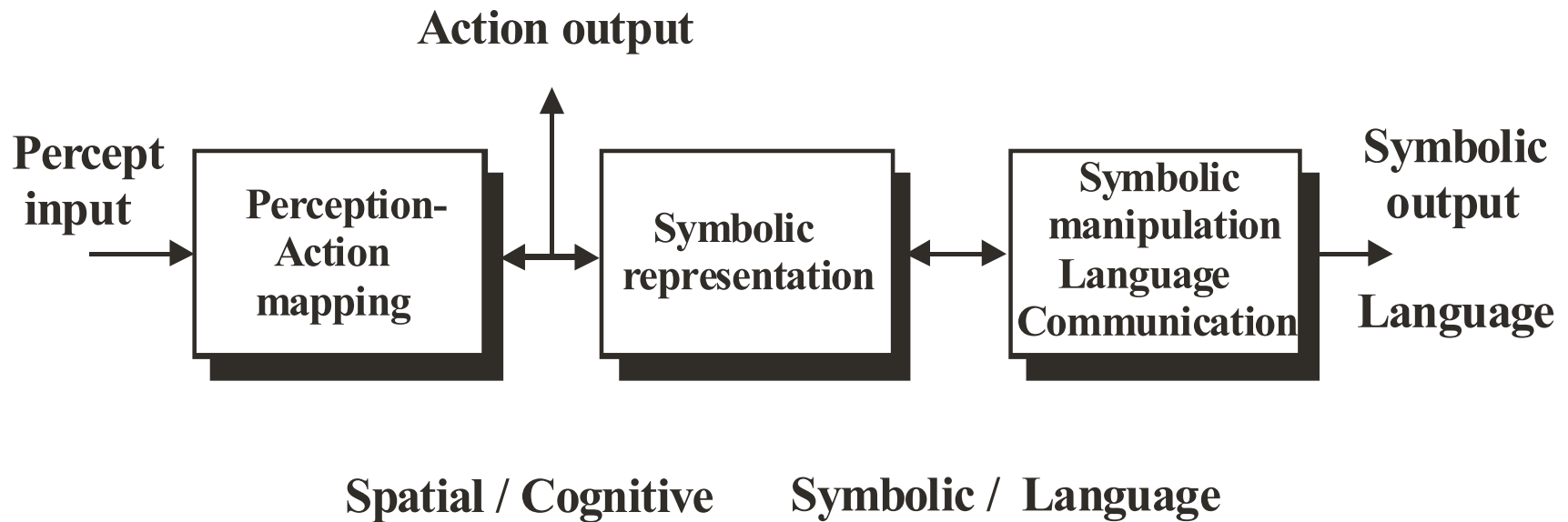
Why active learning?

Act-perceive-learn cycle

- Only features that change are related to the action or state change
- The action or state space is much less complex than the percept space
- Does not require consciousness (other forms of learning do)



Extended Cognitive Structure



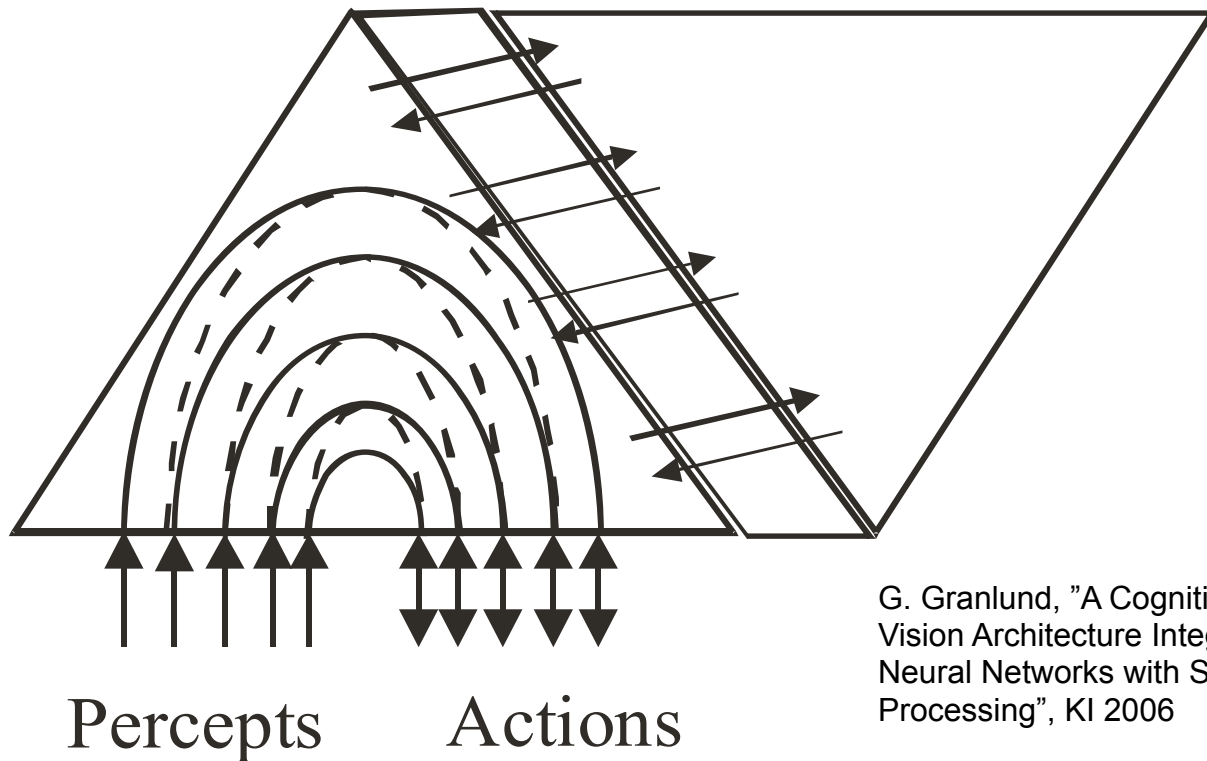
G. Granlund, "A Cognitive Vision Architecture Integrating Neural Networks with Symbolic Processing", KI 2006



Pyramid version

Continuous

Symbolic



G. Granlund, "A Cognitive Vision Architecture Integrating Neural Networks with Symbolic Processing", KI 2006



Summary

- Biological vision systems are not monolithic, but a collection of **visual behaviours**
- Visual perception is done in **cortical maps**, for e.g. colour, edges, and faces
- Much of visual learning is **active**, and pre-conscious