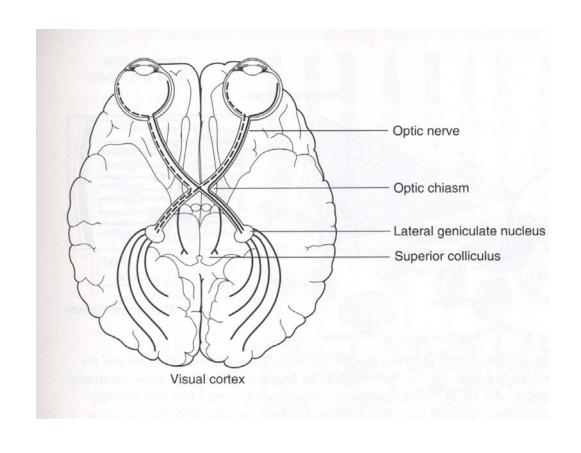
# Lateral Geniculate Nucleus (LGN)

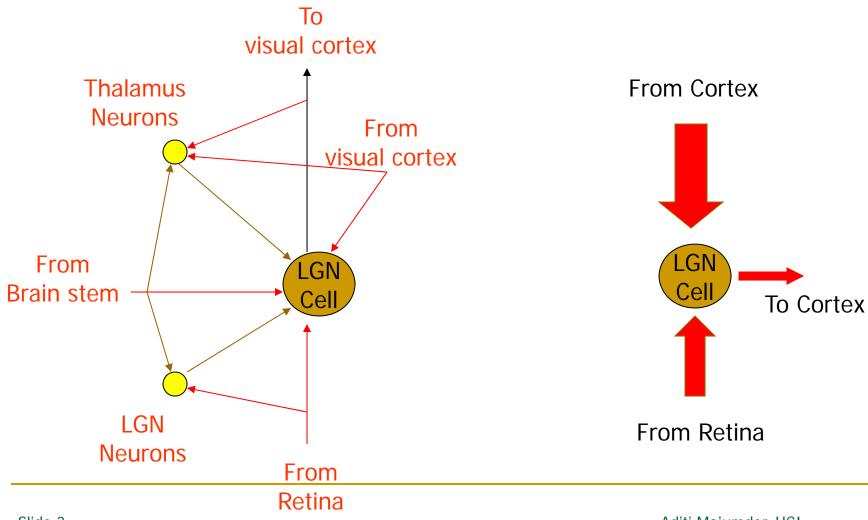
#### What happens beyond the retina?

- What happens in
  - Lateral Geniculate Nucleus (LGN)-90% flow
  - Visual cortex
- Information Flow
  - Superior colliculus
    - 10% flow



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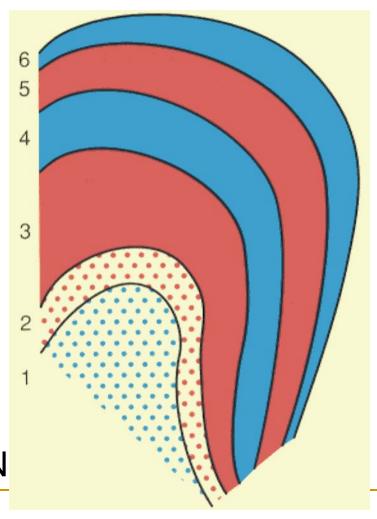
#### Information flow in LGN



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# Information Organization in LGN

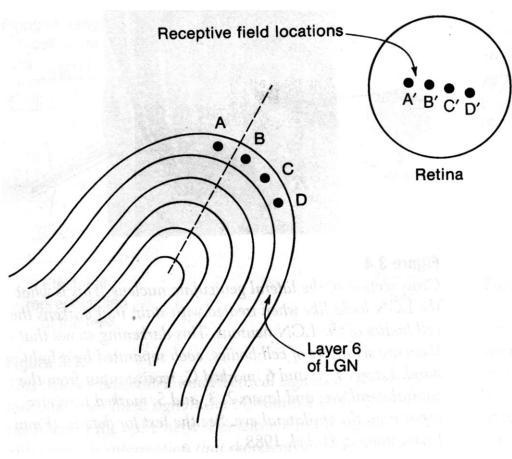
- Bilateral structure with six layers
- 1 million neurons in total
- Each layer receives signal from one eye
- Layer 2,3,5 receives from ipsilateral eye
- Layer 1,4,6 receives from contralateral eye
- Each eye send half information to each side LGN



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## Retinotopic Map

- Each location in LGN maps to a location on retina
- Receptive fields of neurons adjacent to each other in LGN have adjacent receptive fields in retina
- All the neurons on the same column across layers are sensitive to same area on the retina



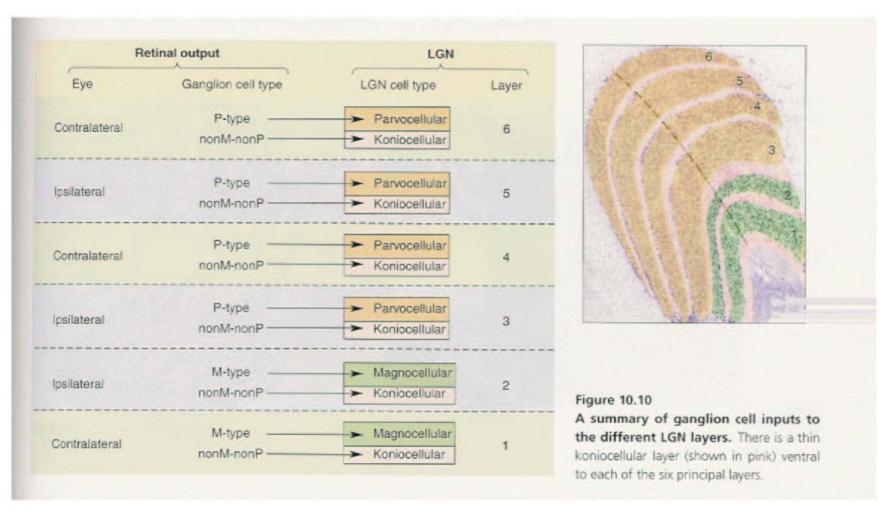
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# Organization by Ganglion Cells

- P-cells (parvocellular)
  - Small medium sized cell body
  - Reaches layers 3,4,5,6
  - Responsible for color, fine textures, patterns and details vision
- M-cells (magnocellular)
  - Larger cell bodies
  - Reaches layers 1,2
  - Responsible for motion detection
- K-cells (koniocellular)
  - Largest cell bodies
  - Reaches all the six layers

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# Summary of LGN Organization



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# Visual Striate Cortex

# Information Processing

- 250 million neurons
- Process the incoming information from LGN
  - Make it clearer
- Neurons are specialized to respond to (feature detectors)
  - Orientation
  - Spatial Frequency
  - Length
  - Corners
  - Motion

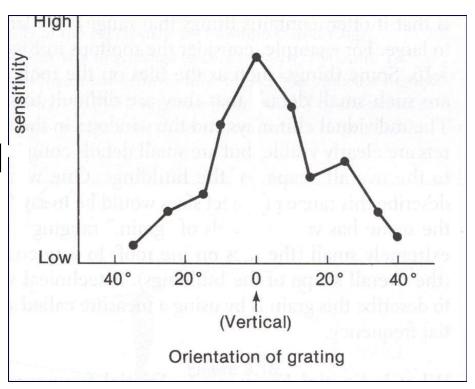
# Three types of Cells

- Simple Cells
- Complex Cells
- End Stopped Cells

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# Simple Cells

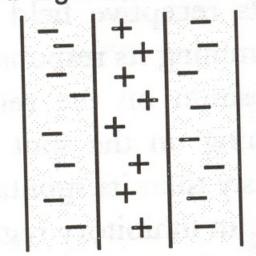
- Orientation sensitive
- Cylindrical shape excitatory center and inhibitory surround
  - Edge detectors
- Four different kinds

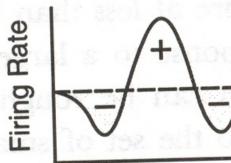


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# Four kinds of Edge Detectors

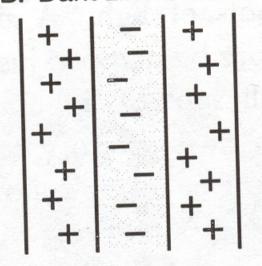


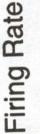


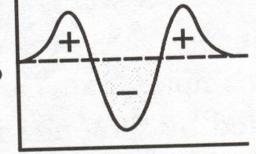


Horizontal Position

#### B. Dark Line Detector



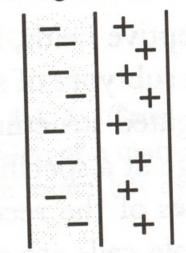


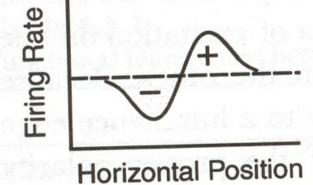


Horizontal Position

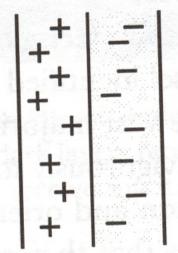
# Four kinds of Edge Detectors

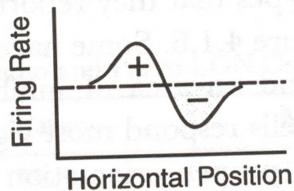
#### C. Dark-to-light Edge Detector



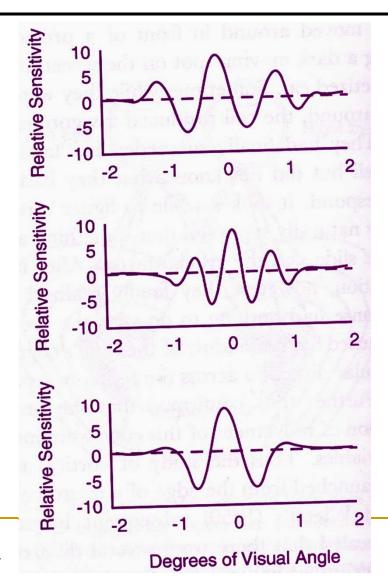


D. Light-to-dark Edge Detector





## May have complex receptive field



- Secondary lobes
- Adjacent to primary lobes

# Selective Adaptation

- When we view a certain state of a property
  - Initial phase : Neurons sensitive to it have increased firing
  - Later phase: This firing reduces and we become insensitive to this property
- Adapt selectively towards one particular state

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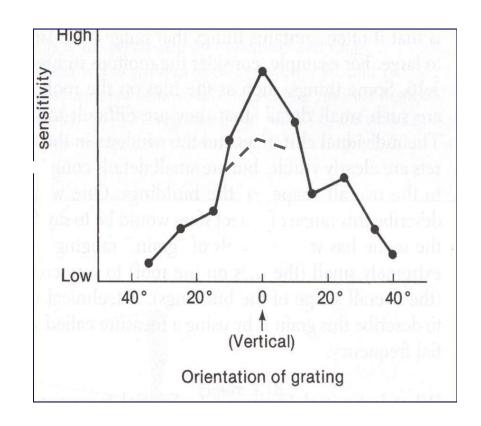
#### Example Experiment: Orientation

- Measure sensitivity to all different orientation
- Adapt to one orientation
- Measure sensitivity to all different orientation again
- Find the change in the sensitivity

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# Selective Adaptation

If stimulated by the same stimulus for a long time, sensitivity to the stimulus decreases



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# Designing the Experiment

- Stimulus
  - Grating Pattern
- Sensitivity Measure
  - Contrast threshold
- Contrast
  - Define as the amplitude/mean of grating
  - Threshold : amount of contrast required to detect grating
  - Sensitivity: 1/threshold

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# Complex Cells

- Orientation Sensitive
- Non-linear
  - Never respond to stationary spot
  - Difficult to find their receptive field
- Motion Sensitivity
  - Responds to moving lines
  - Often, depends on the direction of motion

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# Complex Cells

- Position Insensitivity
  - Does not matter is the position of stimulus changes slightly
- Spatial Extension
  - Larger receptive fields than the simple cells

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# End Stopped Cells

- Far more specific
- Respond to moving lines and corners
  - Of specific length
  - Of specific orientation
  - Of specific size
- Probably hypercomplex version of the simple and complex cells

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#### What this shows...

- Our eye is very sensitive to edges, corners
- Any kind of features

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# Organization

- Retinotopic Map
- Cortical Magnification
  - Fovea is 0.01% of retina
  - Retinotopic map of fovea is 8-10% of cortex

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# Cortical Magnification

- Density of receptors and ganglion in retina is very mismatched
  - In fovea about 50,000 ganglion cells per sq. mm.
  - In periphery about 1000 ganglion cell per sq. mm.
- Density of neurons from fovea and periphery is close to uniform

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# Cortical Magnification

- Foveal input allocated extra cortical neurons
- Three to six times more than those at periphery
- One of the factors for higher acuity in the fovea

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#### Location Columns

- Retinotopic Map
- Same region on retina along the depth
- Adjacent regions in the retina correspond to adjacent columns

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#### Orientation Columns

- Orientation column perpendicular to retinotopic columns
- Same columns have similar orientation preference
- Orientation preference changed across columns continually
- For every 1mm region, the entire range of orientations were covered

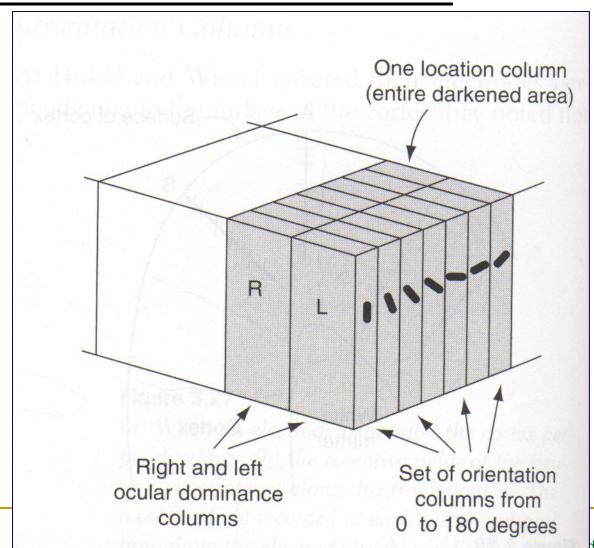
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#### Ocular dominance columns

- Each columnar region is more sensitive to one eye
- The dominant eye alternates between the columnar regions

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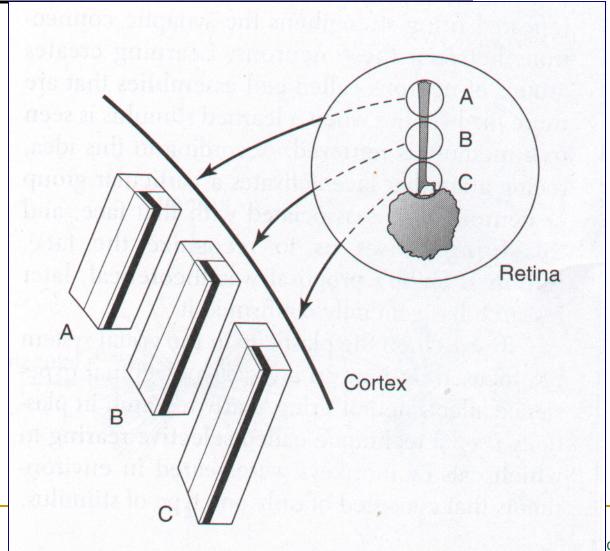
# Hypercolumn View of the Cortex



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## Information Processing by Cortex



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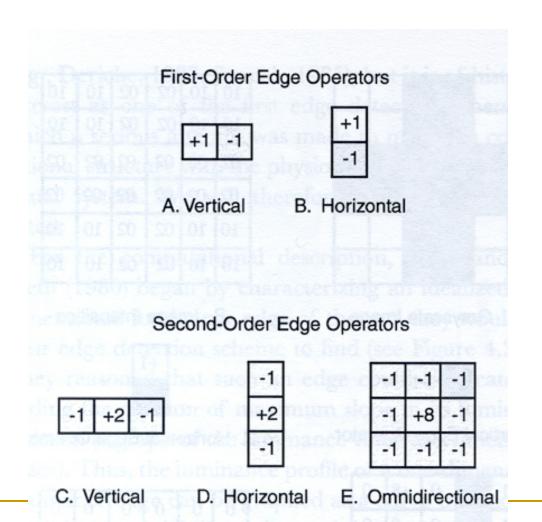
### Development of Receptive Fields

- Is it there from birth?
- How much is the development dependent on learning?
- Experiments with visually deprived kittens

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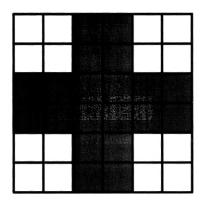
## EDGE DETECTION

# Edge Detectors



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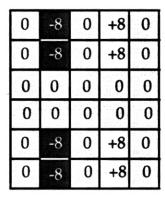
#### Convolution & Feature Detection



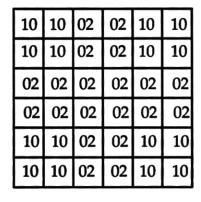
A. Grayscale Image



C. Vertical Edge Operator



E. Convolution of Image with Vertical Edge Operator



B. Image Intensities

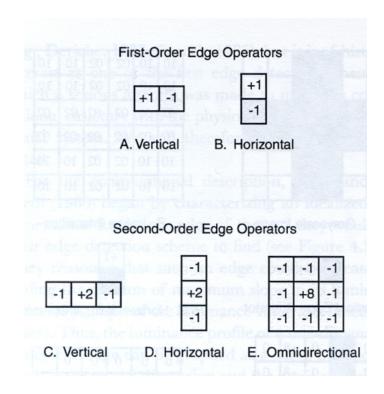


D. Horizontal Edge Operator

0	0	0	0	0	0
+8	+8	0	0	+8	+8
0	0	0	0	0	0
-8	-8	0	0	-8	-8
0	0	0	0	0	0

# Edge Detectors

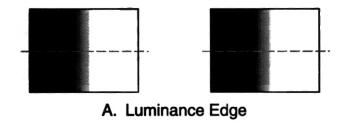
- Finds the slope
  - First derivative
  - Direction dependent
  - Need many edge detectors for all orientation
- Second order derivatives
  - Marr Hildreth Method

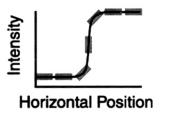


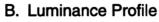
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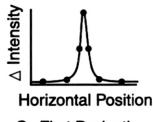
# Zero Crossing Curvature

- Maxima of derivative
  - Causes a zero at second derivative
- Symmetric about the zero
- All direction edges can be detected by this zero crossing

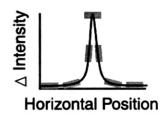




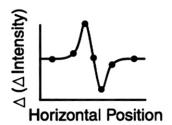




C. First Derivative of Luminance Edge

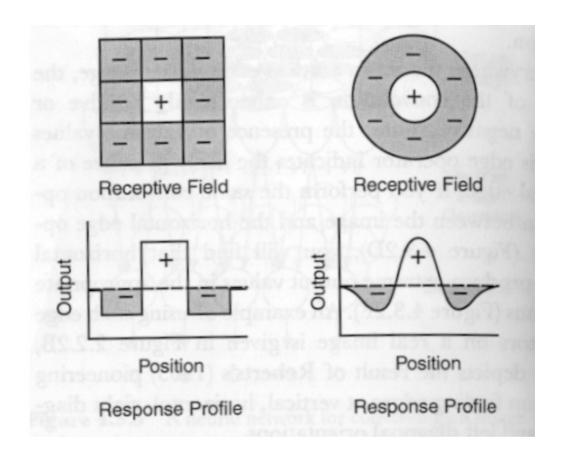


D. First Derivative of Luminance Edge



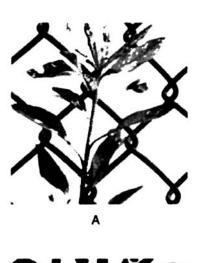
E. Second Derivative of Luminance Edge

# Similarity with Receptive Fields



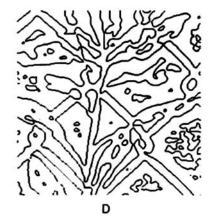
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## Results of the Algorithm







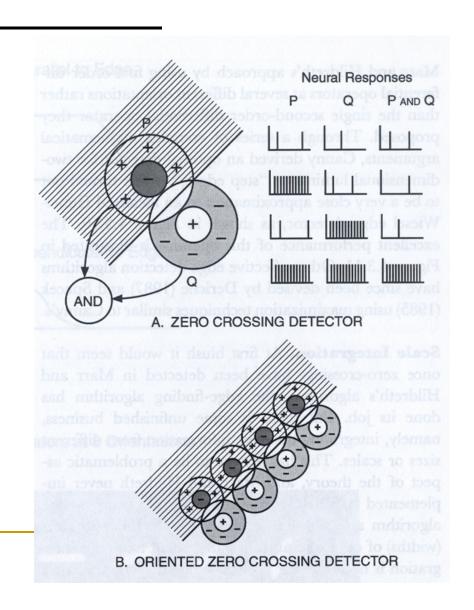


- A. The image
- Image after convolution
- c. Segmented convolved image
- Edge detected image

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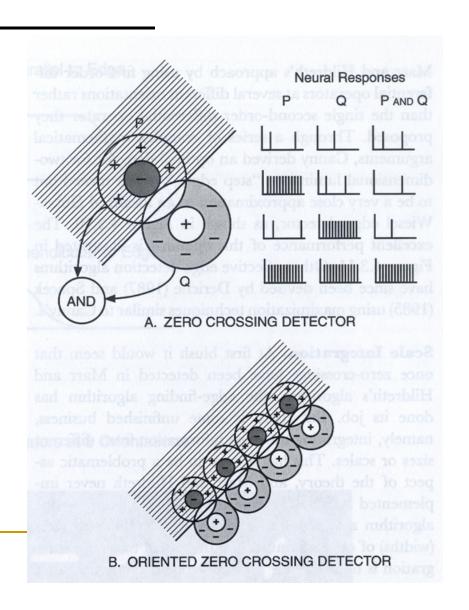
#### Neural Model

- Three cells
  - Convolution
  - Maxima detection
  - Zero detectors
  - Aligned zero detectors to form edge detectors



#### Neural Model

- Three cells
  - Convolution
    - Lateral Inhibition
  - Maxima detection
    - Simple Cortical Cells
  - Zero detectors
    - Complex Cortical Cells
  - Aligned zero detectors to form edge detectors

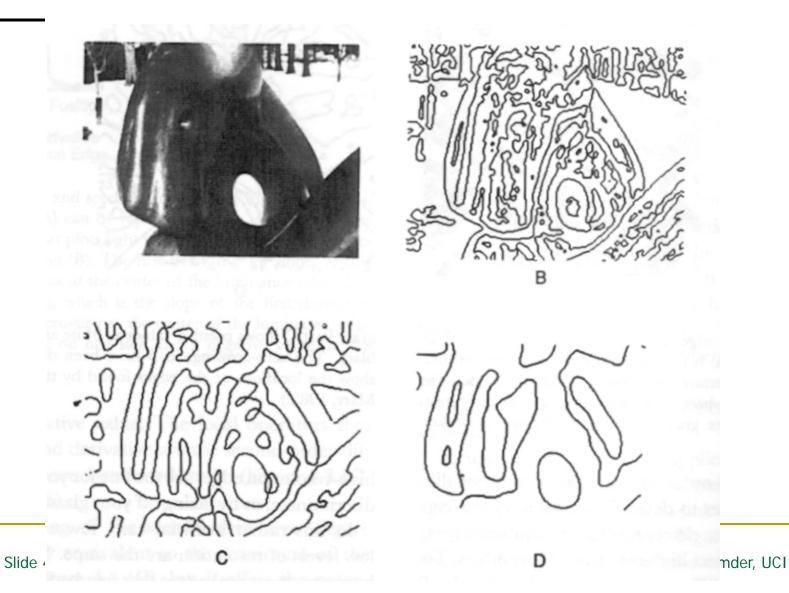


## Scaling Problem

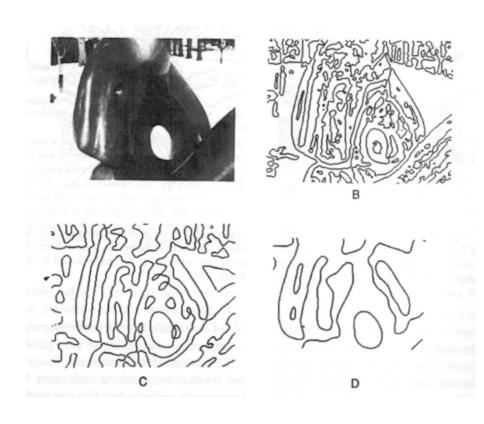
- Can occur in different scales or sizes
  - Some have the transition over a broader region
  - Some over a smaller region
  - Edges nevertheless and has to be detected
- Edge detection are done at several levels
  - Image is sub sampled
    - Reduces information content
  - Then edge is detected

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## Scaling Problem



#### Scaling Problem



- Edges in coarser level do not disappear in finer levels
- New edges are added
- Coarser level edges are most important
- Advances like a hierarchy

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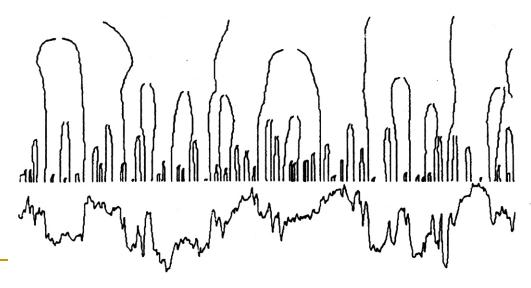
## Scale Integration

- Different resolution images in different levels
- How do we know where the coarser level edges are in the finer edge detected image
- Seems very complex yet eye does it easily

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## Witkin's Explanation

- If we do a continuous subsampling
  - Not possible in digital domain
- Edges are retained, new edges are added with refinement



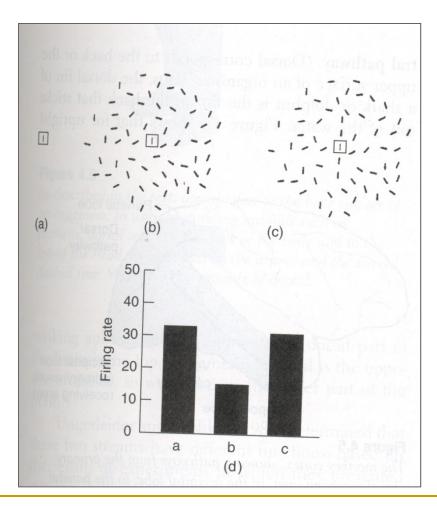
# HIGHER-LEVEL VISUAL PROCESSING

#### Extrastriate Cortex

- Hubel and Weisl won Noble prize for their discovery of the cortical cells
- By 1970s, found that other regions of the brain are also involved in vision.
- Cells that respond to far more complex stimuli

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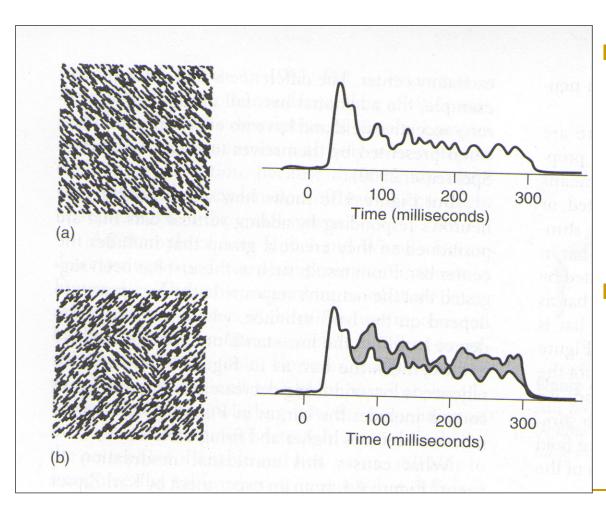
#### Contextual Modulation



- Stimulation can be changed by changing their context
- Salience: Degree to which things stand out

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## Physiological Explanation

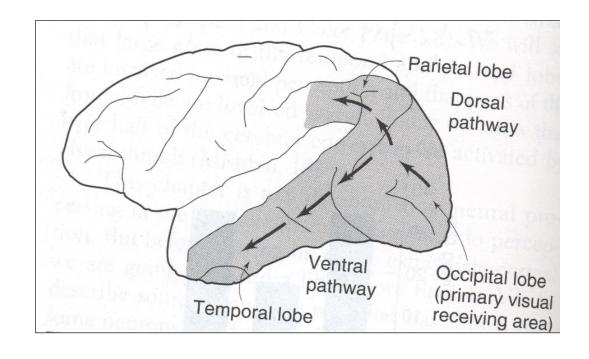


- Notice 80ms initial quiet
  - Time required to process salience
- Adverse effect on selective adaptation

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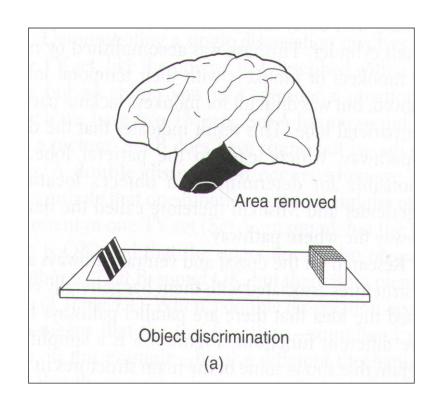
#### Processing Streams

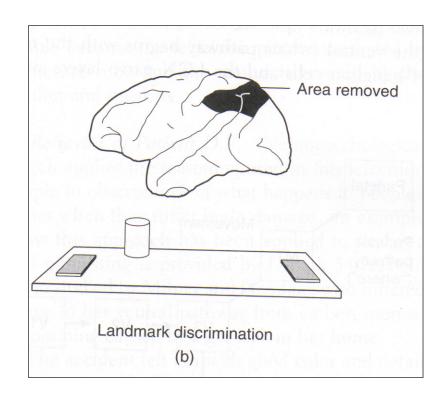
- Dorsal Pathway
  - Parietal Lobe
  - 'Where'
  - Location and Action
- Ventral Pathway
  - Temporal Lobe
  - 'What'
  - Object Discrimination



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# Supporting Experiments

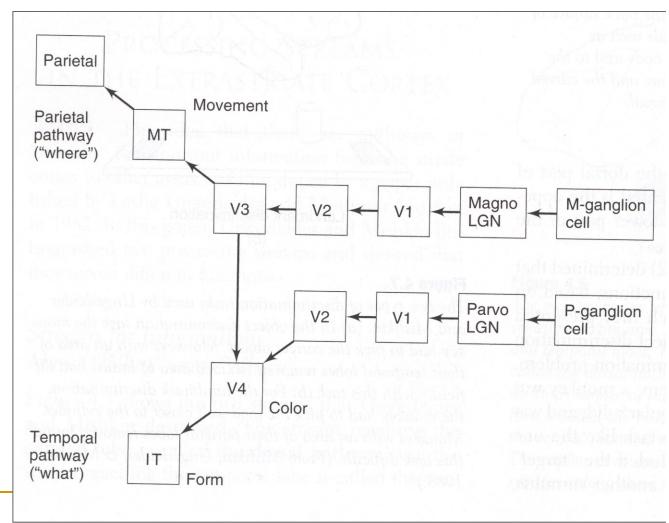




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## The Whole Pathway

- Parallel Path
- Not independent
- P-cells to ventral
- M-cells to parietal



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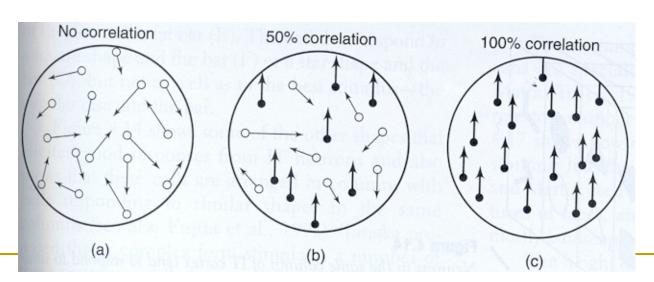
#### Modular Neurons

- Neurons in MT and IT
- Process very specific information
- Experiment of motion correlation

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#### Experiment with motion correlation

- If MT present, can detect as small as 1-2% correlation
- If MT absent, cannot detect less than 10-20% correlation



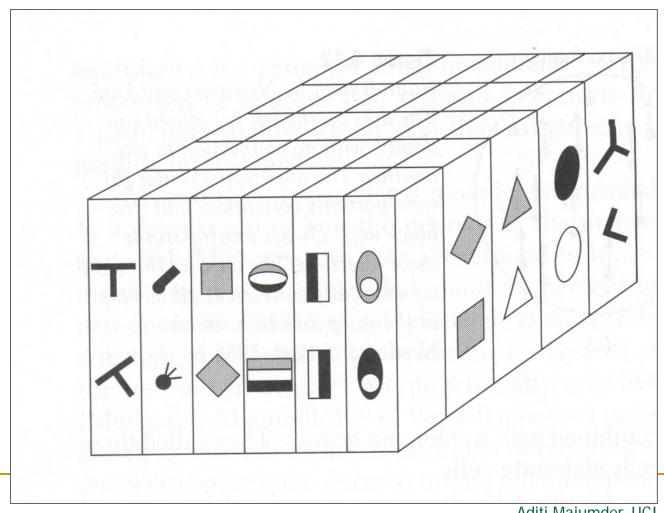
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## Infotemporal cortex

- Primary Cells
  - Respond to simple stimuli
    - Slits, spots, ellipses, squares
- Elaborate Cells
  - Responds to complex stimuli
    - Specific shapes, shapes with color and texture

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# Hypercolumn Again



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#### Neurons respond to faces

- If a body with face is presented as stimuli, they fire
- When the face is covered with paper, they stop firing
- fMRI research with humans
  - Fusiform face area (FFA) or fusiform gyrus
- Prosopagnosia
  - Due to damage to temporal lobe
  - Fusiform gyrus

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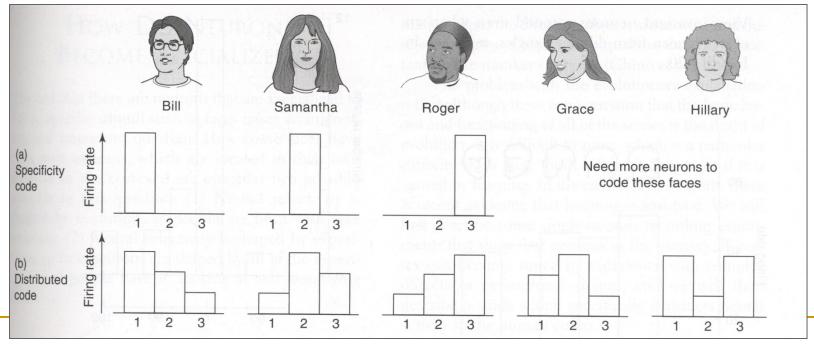
#### The Sensory Code

- Information encoded in the firing of neurons
  - Specificity Coding
    - Every neuron responds to a specific stimuli
  - Distributed Coding
    - Difference in amounts of response in several different neuron creates a pattern that causes identification of specific stimulus
    - To some extend, like number system
    - More number of levels for each neuron, lesser the number of neurons needed for coding

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#### The Sensory Code

- Information encoded in the firing of neurons
  - Specificity Coding
  - Distributed Coding



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## Neurons in IT for Face Recognition

- Size invariant
- Location invariant
- View invariant
- Size specific
- Location specific
- View specific

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#### Role of Attention

- Selectivity of attention
  - Directs our receptors to stimuli
  - Enhances the perception of stimuli

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#### Inattentional Blindness

- Cannot recognize shape of unattended object when attending to some other visual job
- Experiments of shape presented while performing the task of identifying shorter length
- https://www.youtube.com/watch?v=vJG698U2 Mvo

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#### Attentional Blink

#### Attentional Blink

- Inability to attend to a different stimulus within a short period of 500ms
- Demonstrated by rapid serial visual presentation (RSVP)

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## Change Blindness

- ChangeBlindness
  - Inability to detect unattended changes
  - Even when the stimulus is presented slowly









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#### Examples

- https://www.youtube.com/watch?v=uO8wpm 9HSB0
- https://www.youtube.com/watch?v=bh\_9XFz bWV8
- https://www.youtube.com/watch?v=FWSxSQ sspiQ

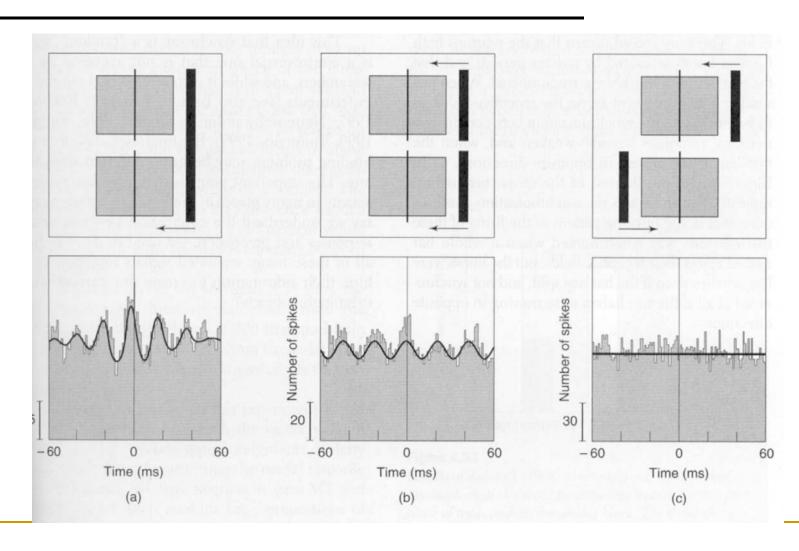
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## Binding Problem

- How do we know it is all from the same object
  Example of a car
- Depends on the synchrony of neuron firing
- If same object, all the different neurons fire synchronously
- Cross correlogram plots of the brain

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# Cross Coreleograms



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