

# CISC 3250 Systems Neuroscience

## Perception



Professor Daniel Leeds  
dleeds@fordham.edu  
JMH 332

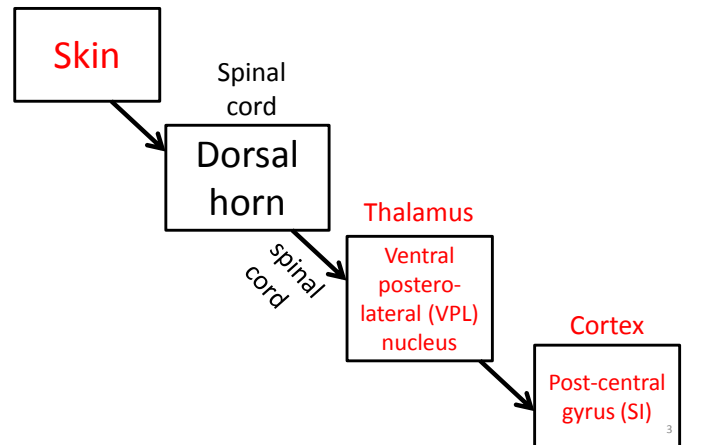
## Pathways to perception in 3 (or fewer) synaptic steps

- 0 Input through sensory organ/tissue
- 1 Synapse onto neurons in spinal cord/brain stem
- 2 Synapse onto neurons in thalamus
- 3 Synapse onto cortical neurons in "primary \_\_\_\_ cortex"
- 4+ Further cortical processing



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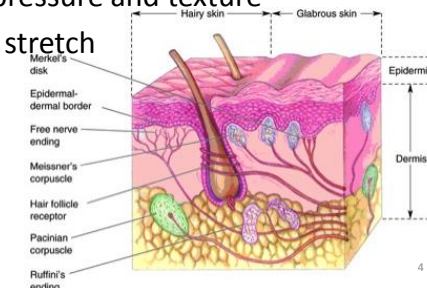
## Touch/"Tactile"



## Touch: Inputs

### Mechanoreceptors in skin

- Pacinian corpuscles – vibrations
- Meissner's corpuscles – light touch
- Merkel's discs – pressure and texture
- Ruffini endings – stretch



## Thalamus – the “relay” station

Region names largely based on location

VPL for somatosensation

**Legend**

- Anterior nuclei
- Medial nuclei
- Lateral nuclei
- LP - Lateral posterior nucleus
- LD - Lateral dorsal nucleus
- VA - Ventral anterior nucleus
- VL - Ventral lateral nucleus
- VP - Ventral posterior nucleus
- VI - Ventral intermediate nucleus
- VPM - Ventral posteromedial
- VPL - Ventral posterolateral

<http://en.wikipedia.org/wiki/File:Thalamus-schematic.svg>

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## Hearing/“Auditory”

Cochlea

Cochlear nerve

Cochlear nucleus (-> Superior olive) -> Inferior colliculus

Brain stem

Thalamus

Medial geniculate nucleus (MGN)

Cortex

Primary auditory cortex (AI)

## Hearing and frequency decomposition

Sound consists of times and frequencies

Time-bound wavelets:

Similar to cochlear neurons

$$w(t) = \frac{2}{\sqrt{3\sigma\pi}^{1/4}} \left( 1 - \left( \frac{t}{\sigma} \right)^2 \right) e^{-\frac{t^2}{2\sigma^2}}$$

“Mexican hat”

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## Common patterns in speech

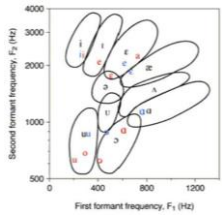
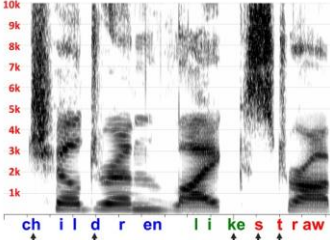
- Vowels (a,e,i,o,u) correspond to steady frequency combinations

- Consonants may be broad-range frequencies, or sweeps

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### More speech pattern

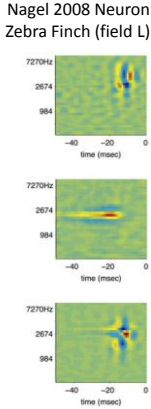
- Speech formant ranges by frequency
- ch, s – long high freq
- d, k, t – broad freq burst
- l, r, n, m – freq slide

### Spectro-temporal receptive fields

AI (primary auditory cortex) neurons selective for patterns in space and time

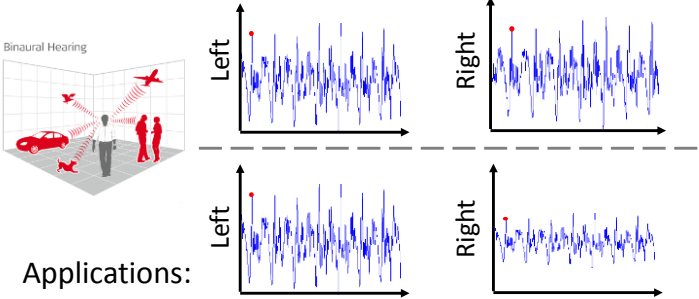
Nagel 2008 Neuron Zebra Finch (field L)



### Binaural hearing

Comparing sounds from left and right

- Time shift and/or Volume Change



Applications:

- Localize sound source
- Distinguish sounds from multiple sources

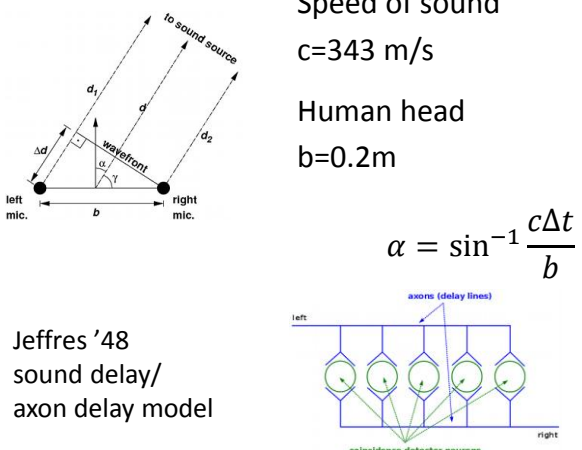
### Math of sound localization

Speed of sound  $c=343$  m/s

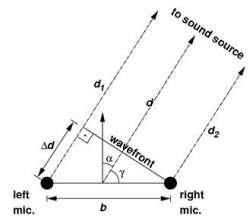
Human head  $b=0.2$ m

$$\alpha = \sin^{-1} \frac{c\Delta t}{b}$$

Jeffres '48 sound delay/axon delay model



## Math of sound localization



Speed of sound

$c=343$  m/s

Human head

$b=0.2$ m

$$\alpha = \sin^{-1} \frac{c\Delta t}{b}$$

Pick direction for comparison

$$\Delta t = \begin{cases} > 0 & \text{rightSound earlier} \\ < 0 & \text{leftSound earlier} \end{cases}$$

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## Sound localization examples

Speed of sound

$c=343$  m/s

Human head

$b=0.2$ m

$$\alpha = \sin^{-1} \frac{c\Delta t}{b}$$

Sound arrives at ears:

R: 324.6ms L: 324.4ms

$$\alpha = \sin^{-1} \frac{343 \times 0.0002}{0.2} = \sin^{-1} 0.343 =$$

**20° to the left**

Sound arrives at ears:

R: 512.5ms L: 513.0ms

(sound comes to left ear first)

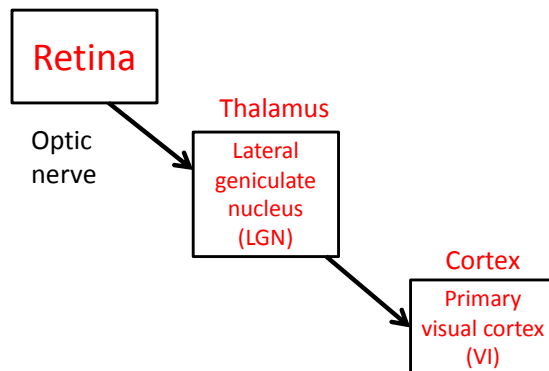
$$\alpha = \sin^{-1} \frac{343 \times 0.0005}{0.2} = \sin^{-1} \frac{0.00176}{0.2} =$$

$\sin^{-1} 0.88 = \mathbf{62^\circ \text{ to the right}}$

(sound comes to right ear first)

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## Seeing/"Visual"



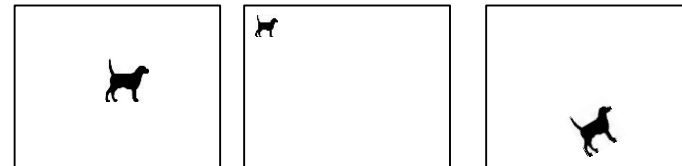
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## Sensitivity to perceptual variations

- V1: Surround-suppression for shifted edges

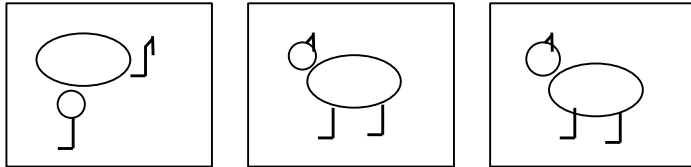


- PFC: Same object detected at diverse locations and scales



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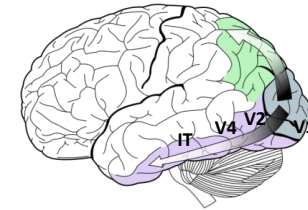
## Selectivity to perceptual variations



- More complex percepts invariant to greater spatial transformations

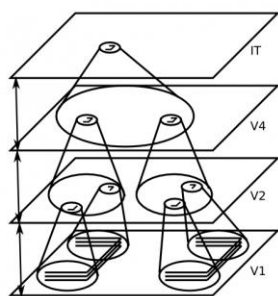
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## HMAX – model of hierarchical vision



- Higher cortical levels cover larger visual spans
- Object recognition invariant to changes in location and orientation

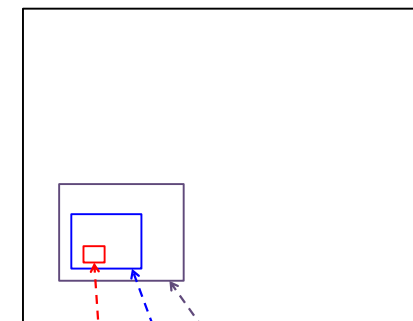
## HMAX – model of hierarchical vision



1. Gabor “filters” (edge detectors)
2. Perform “Max pooling” (semi-invariance over space)
3. Weighted combination of space-invariant edges
4. Further max pooling

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## Higher HMAX layers cover more space




Example coverage  
for layer x neurons

layer 1  
layer 2  
layer 3


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### Functions of HMAX layers

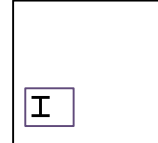
- Odd layers (layer 1, 3, 5, ...) look for specific combinations of lower-level features
- Even layers (layer 2, 4, 6, ...) provide invariance to some feature changes (e.g., shift in position)



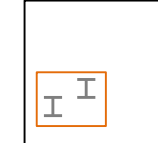
layer 1



layer 2  
Fire for 1+ lines



layer 3



layer 4  
Fire for 1+ Is

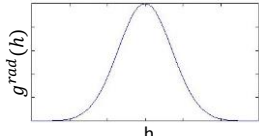
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### Functions of HMAX layers

- Odd layers (layer 1, 3, 5, ...) look for specific combinations of lower-level features

$$h = \sum_j w_j r_j^{in} \quad r^{out} = g^{rad}(h)$$

Radial basis function



- Even layers (layer 2, 4, 6, ...) provide invariance to some feature changes (e.g., shift in position)

$$r^{out} = \max([r_1^{in} \quad r_2^{in} \quad \dots \quad r_j^{in}])$$

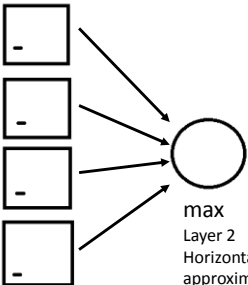
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### Detecting triangles: layer 2

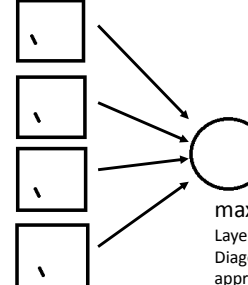
Neuron outputs 1 if desired image viewed, otherwise 0

Layer 1: Specific edge at specific location

Layer 2: Specific edge at slightly varied locations



max  
Layer 2  
Horizontal edge at approximate location



max  
Layer 2  
Diagonal edge at approximate location

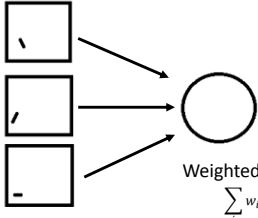
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### Detecting triangles: layer 3

Neuron outputs 1 if desired image viewed, otherwise 0

Layer 2: Specific edge at slightly varied locations

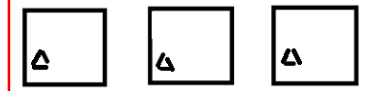
Layer 3: Combination of edges



Layer 2

Weighted sum  
 $\sum_i w_i r_i$

Layer 3  
Triangle centered at fixed location



Accepted stimuli in layer 3 neuron

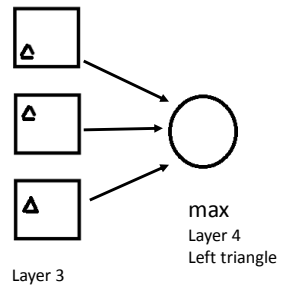
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## Detecting triangles: layer 4

Neuron outputs 1 if desired image viewed, otherwise 0

Layer 3: Combination of edges

Layer 4: Triangle on the left



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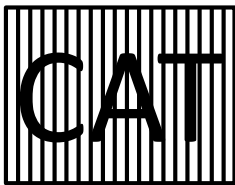
## Visual attention

- Emphasize details currently of interest



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## Attention when percepts overlap



Cocktail party problem



$$h_i = \sum_i w_i r_i^{att} r_i^{in}$$

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## Modulating inputs through multiplication

Algorithm: "Sigma-Pi Node"

- Multiply rates to modulate each input
- Sum to compute output rate

$$h_i = \sum_i w_i a_i^{in} r_i^{in}$$



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### Attention when percepts overlap

**Attention a**

Ignore vertical edges:  $a_v=0$

Pay attention to all other edges:  $a_{\cdot}=a_{/}=a_{\setminus}=1$

**Weights w**

H-detector looks for | and -  $w_v=w_{\cdot}=1$   $w_{/}=w_{\setminus}=-1$

A-detectors looks for /, \, -  $w_{\cdot}=w_{/}=w_{\setminus}=1$   $w_v=-1$

**Rate r**

If feature present: 1

If feature not present: 0

In this example, |, -, /, \ present

$w_v=-1$  for H

H detector

$\sum_i w_i a_i^{in} r_i^{in} = 0$

$w_v=-1$  for A

A detector

$\sum_i w_i a_i^{in} r_i^{in} = 3$

### Attention when percepts overlap

Can attend to one of two voices (e.g., high-pitched voice or low-pitched voice)

### Modulating inputs through multiplication

**Algorithm: "Sigma-Pi Node"**

- Multiply rates to modulate each input
- Sum to compute output rate

$$h_i = \sum_j w_j a_j^{in} r_j^{in}$$

- $a_j^{in}$  - attention input
- $a_j^{in} = \sum_k r_{jk}^{att}$  - can sum over multiple attention inputs

### Dynamic synaptic reweighting

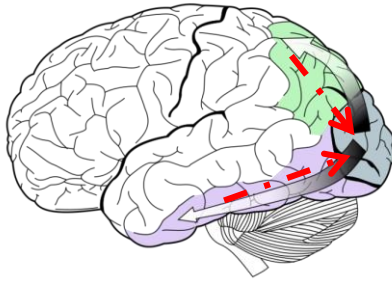
Voltage-dependent NT-receptors (e.g., NMDA):

1. Other nearby receptor decreases voltage
2. Voltage dependent receptor detects NTs



## Complexity of cortical networks

- *Feedback*: connections in both directions along cortical “pathways”



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[http://en.wikipedia.org/wiki/File:Ventral-dorsal\\_streams.svg](http://en.wikipedia.org/wiki/File:Ventral-dorsal_streams.svg)

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