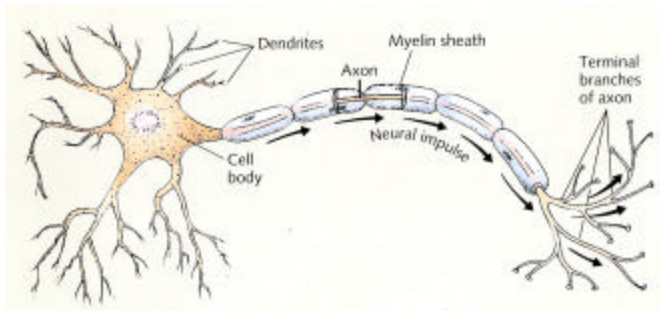


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Week 4, Lecture 2: Neural Computation

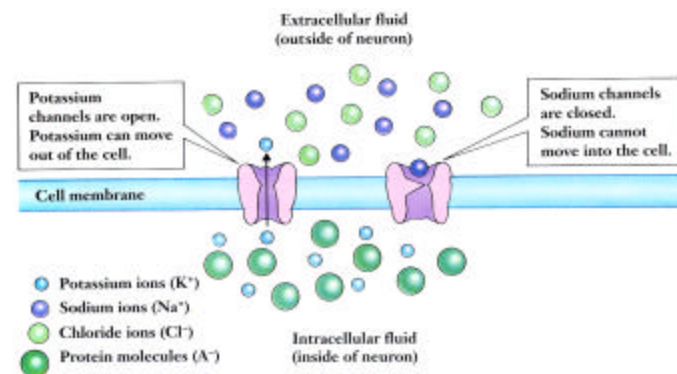
Neurons and Neural Computation

Neurons and Neural Computation

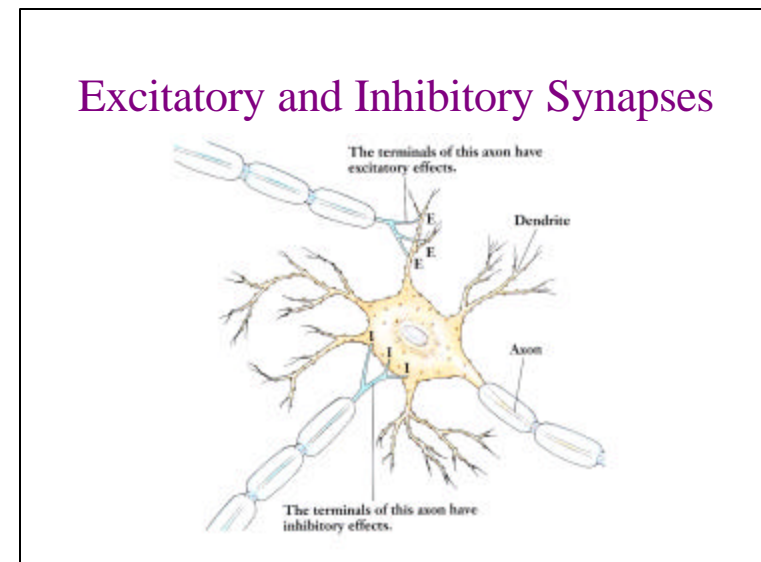
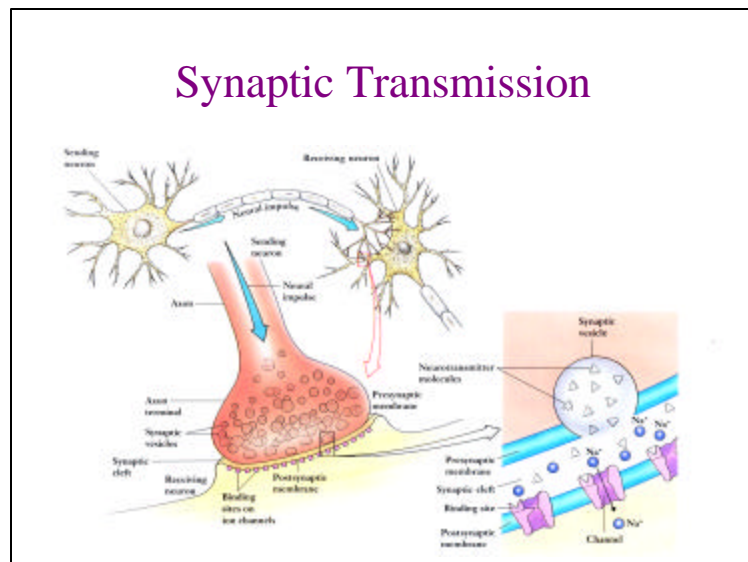
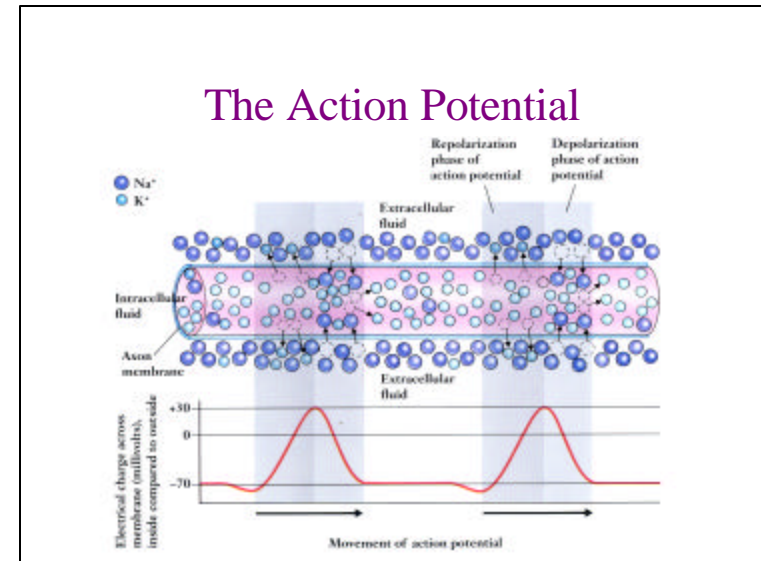
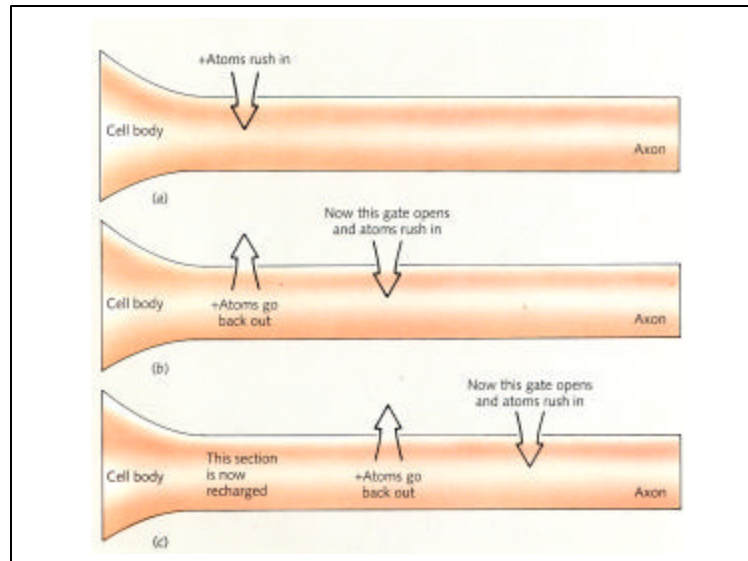
- I. Importance of Neural Computation
 - Neuroscience: What brain tissue does.
 - Computational theory of mind: Implementing elementary information processes.



The Resting Potential

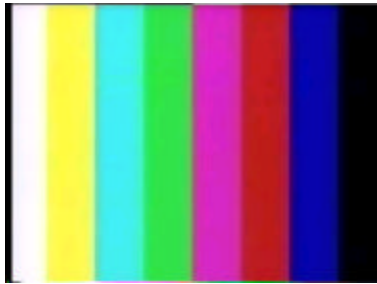


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Feature Detectors

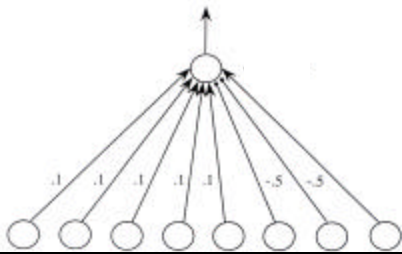


Neural Computation

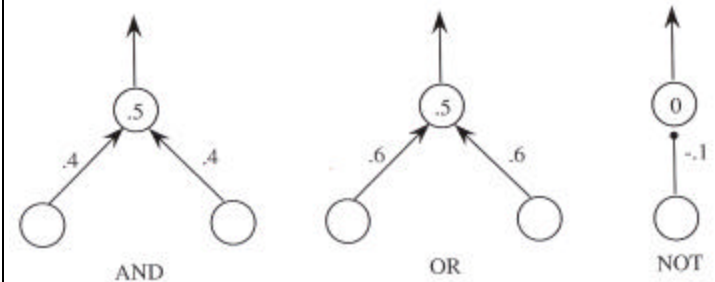
- Computing logical functions with neurons.
Kosher =
{[Chews its cud] AND [Has cloven hooves]}
OR
{[Has fins] AND [Has scales]}

Neural Computation

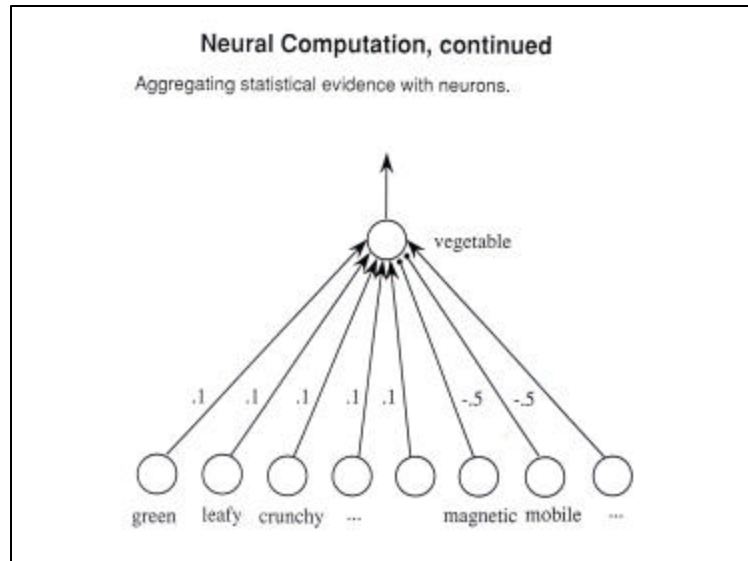
- Multiply each input signal by the "weight" (strength) of the synapse.
- Sum the weighted signals.
- If they exceed the cell's threshold, fire.



Building Logic Gates out of Neurons



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Local vs. Distributed Representations

- Local representation: “grandmother cell,” “yellow Volkswagen detector”
- Distributed representation; “auto-associator”:

A diagram showing a network of eight input nodes at the top, each connected to a corresponding output node at the bottom. The connections are dense and overlapping, representing a distributed representation.

Pattern completion by auto-associators:

A diagram showing the word "MILK" in a stylized font. The letters are partially obscured by a dark, irregular shape, illustrating the concept of pattern completion by auto-associators.

Learning in Neural Networks

- **Neural Computation:**
 - Multiply each input signal by the "weight" (strength) of the synapse.
 - Sum the weighted signals.
 - If they exceed the cell's threshold, fire.
- **Neural Learning:**
 - Change the weights of synapses.
 - Change the thresholds of cells.

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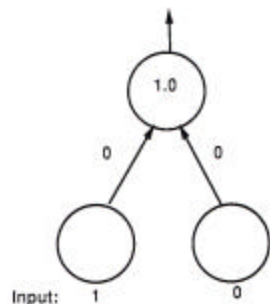
Learning in Neural Networks, continued

- Real but simple example of learning in a neural network: Aplysia (sea snail). See textbook.
- More complex but still hypothetical form of neural learning:

Perceptron Learning Procedure

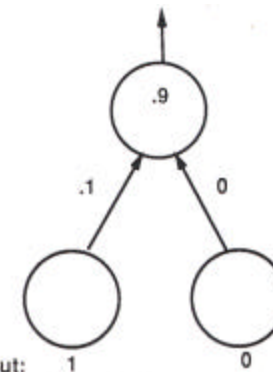
- Compare current output to correct output (from "teacher").
- If too low, increase weights for active inputs, and decrease threshold.
- If too high, decrease weights for active inputs, and increase threshold.

How a two-layer network can learn "OR" with the perceptron learning procedure.



Input: 1 0
Correct Output: 1
Actual Output: $(1 \times 0) + (0 \times 0) = 0$. $0 < 1.0$. Therefore 0.

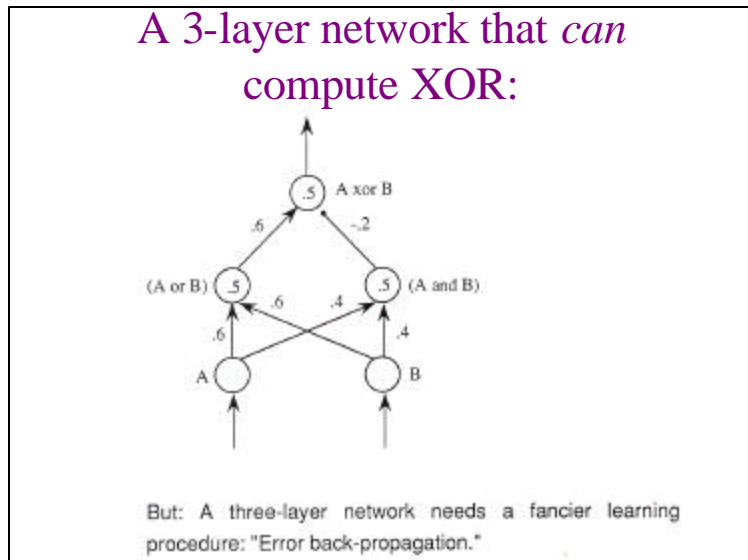
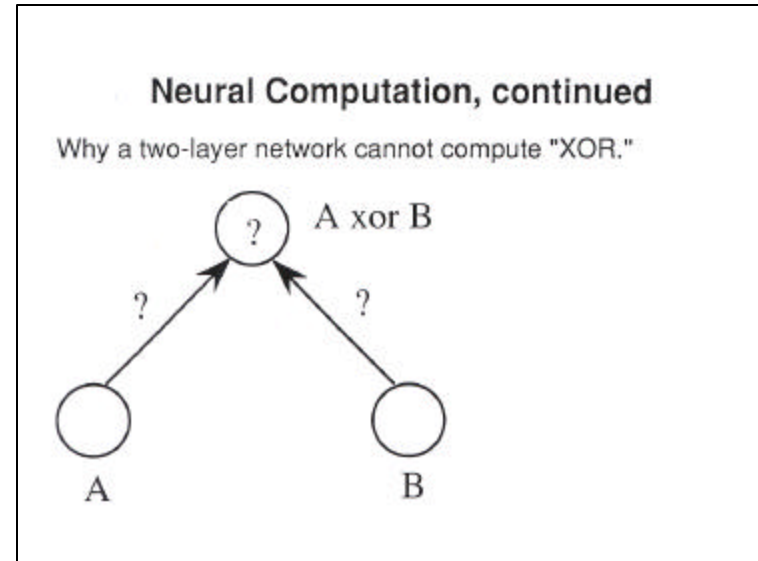
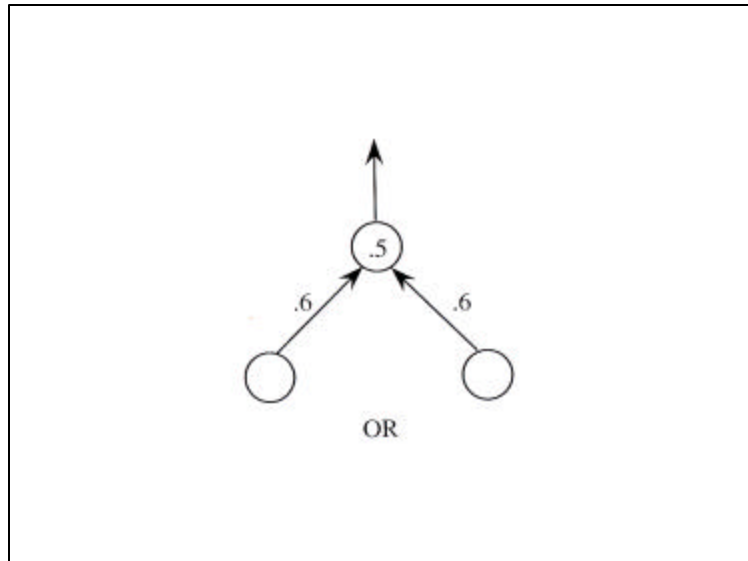
Too small.
Increase first weight by .1.
Leave second weight alone.
Decrease threshold by .1.



Input: 1 0
Correct Output: 1
Actual Output: $(1 \times .1) + (0 \times 0) = .1$. $.1 < .9$ Therefore 0.

Too small.
Increase first weight by .1.
Leave second weight alone.
Decrease threshold by .1.

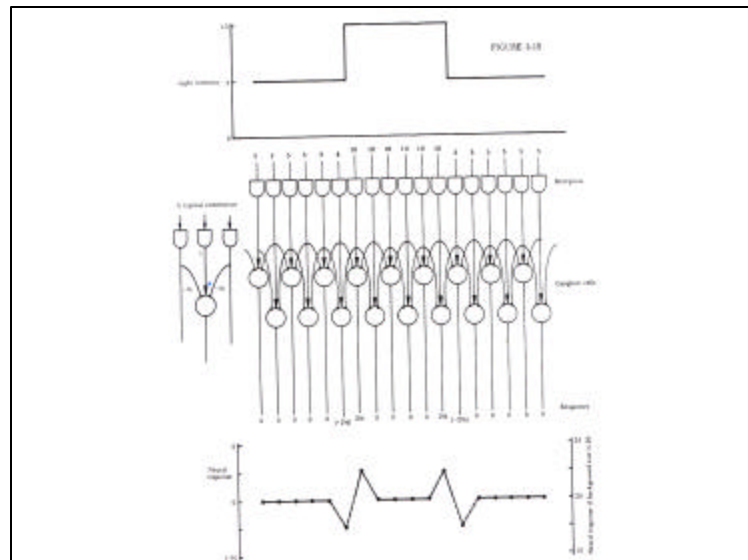
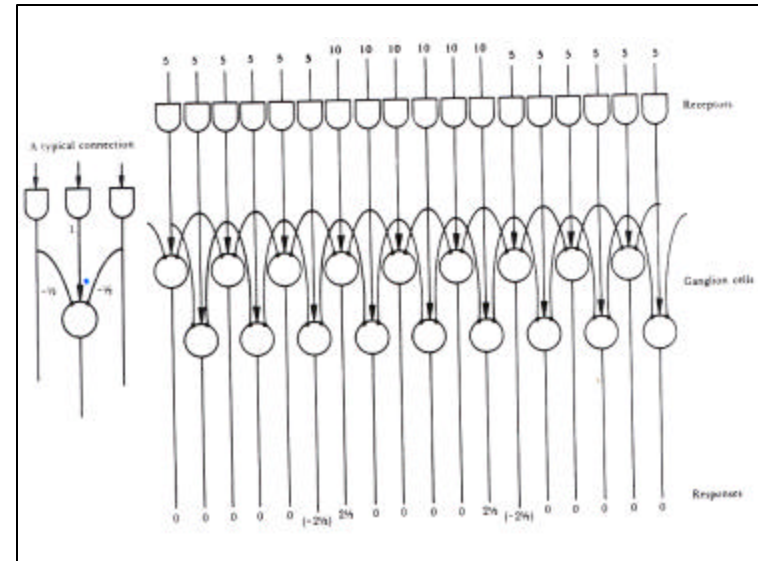
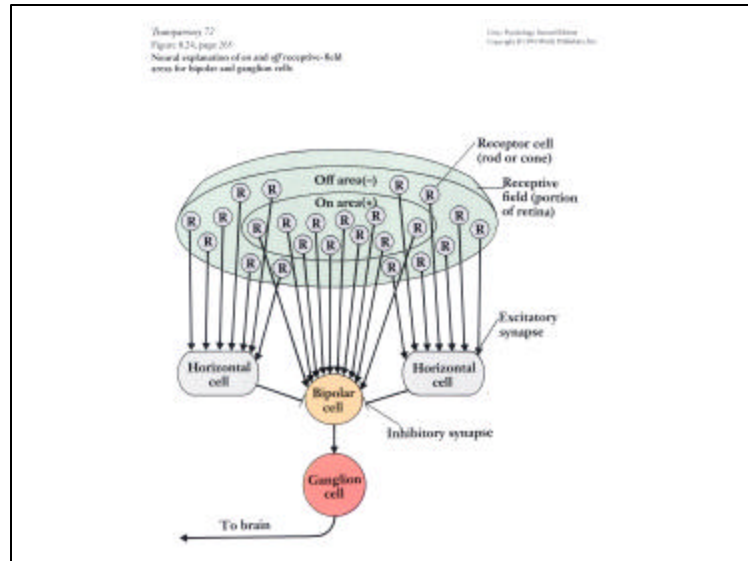
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Relating Neural Networks to Psychology

- Lateral inhibition: Turn on your own output; turn down your neighbor's output.

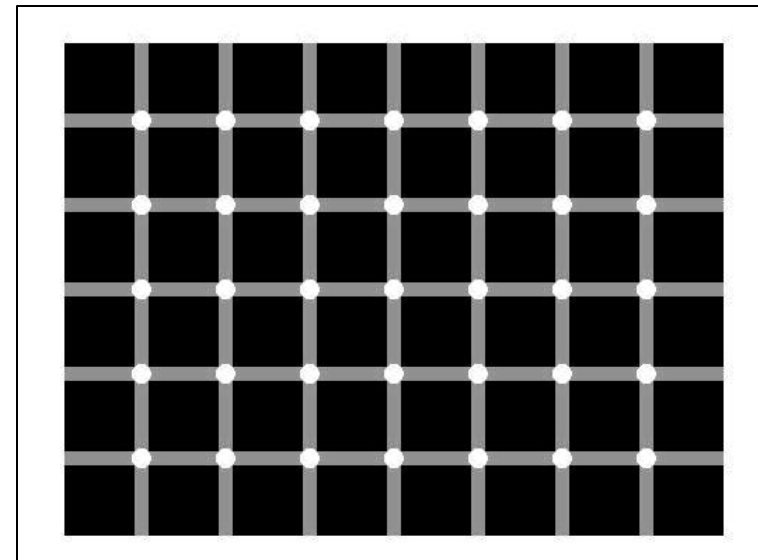
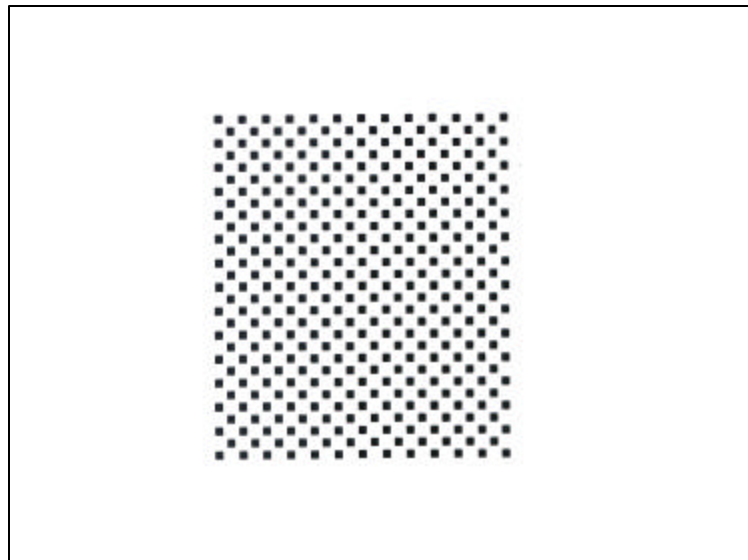
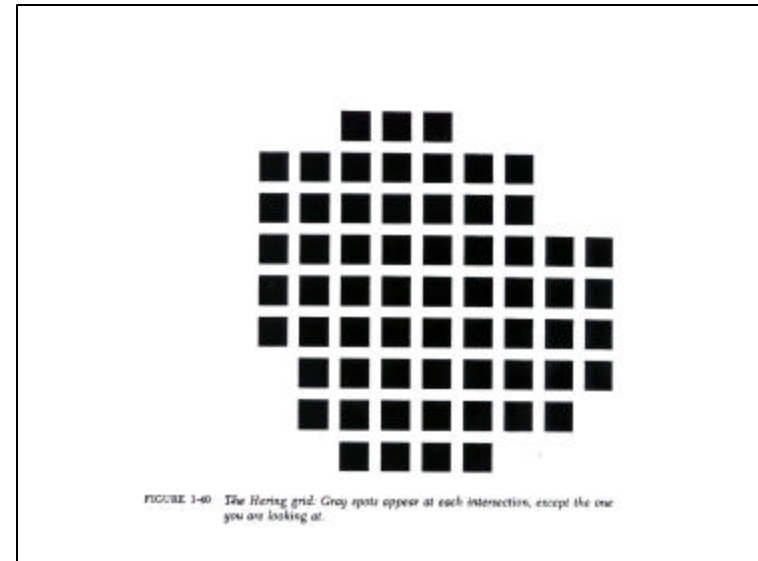
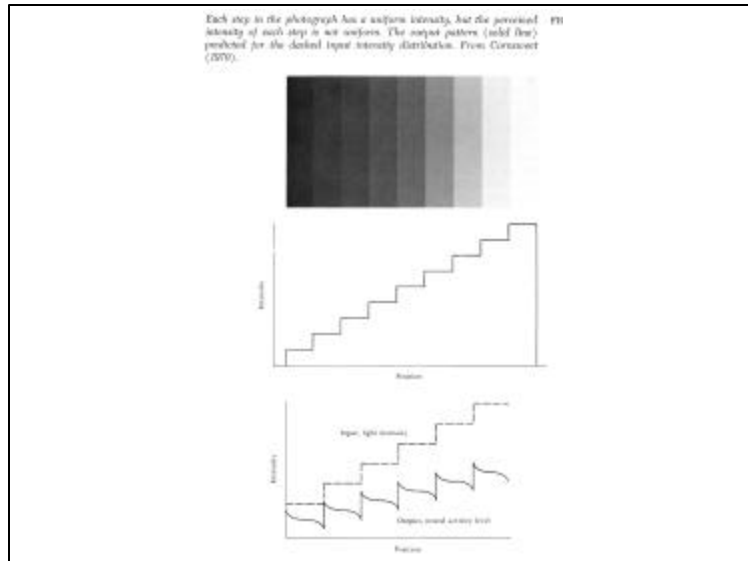
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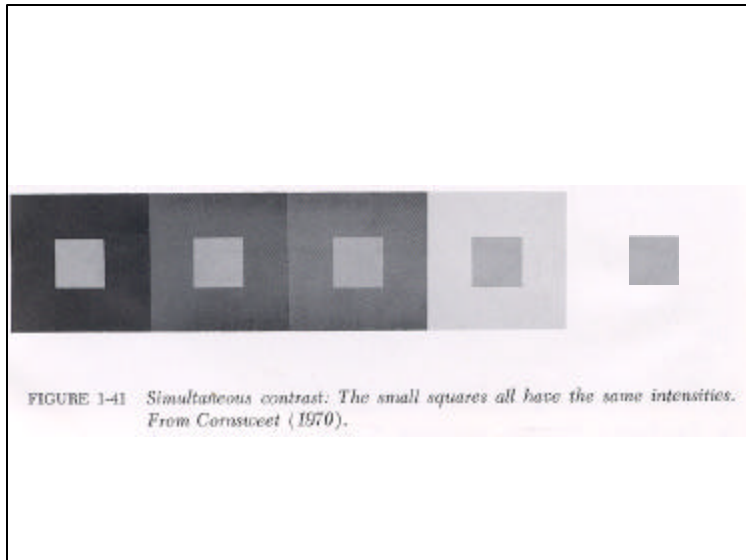
Lateral inhibition as an explanation for:

- Mach bands
- The Hering grid
- Simultaneous contrast

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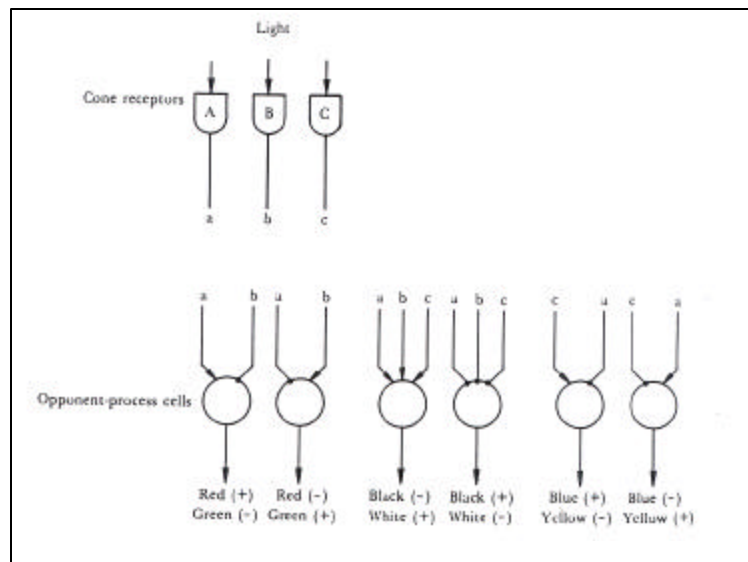


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Relating Neural Networks to Psychology, continued

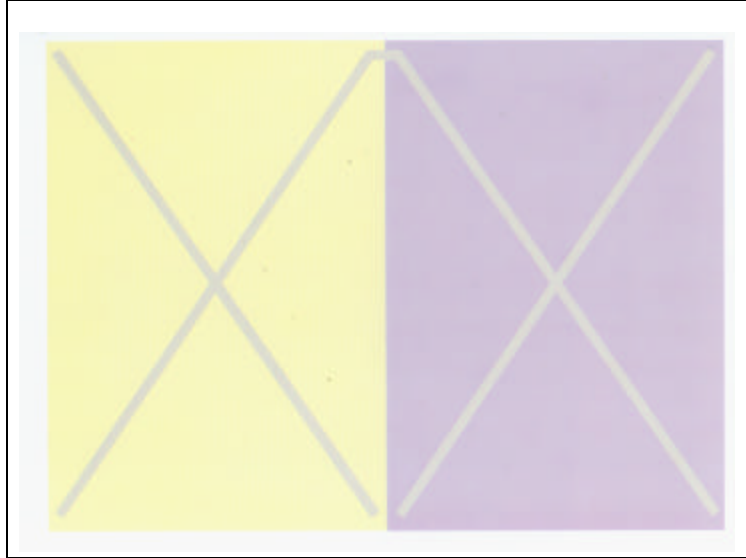
- **Opponent process circuits:**
 - Two inputs to one cell, from opposite kinds of stimuli (red/green, dark/light, move up/down, etc.)
 - A signal for one perceptual quality excites an output; a signal for the complementary quality inhibits the output.
 - The level of activity of the output (excited or inhibited relative to resting level) determines the perceived quality.



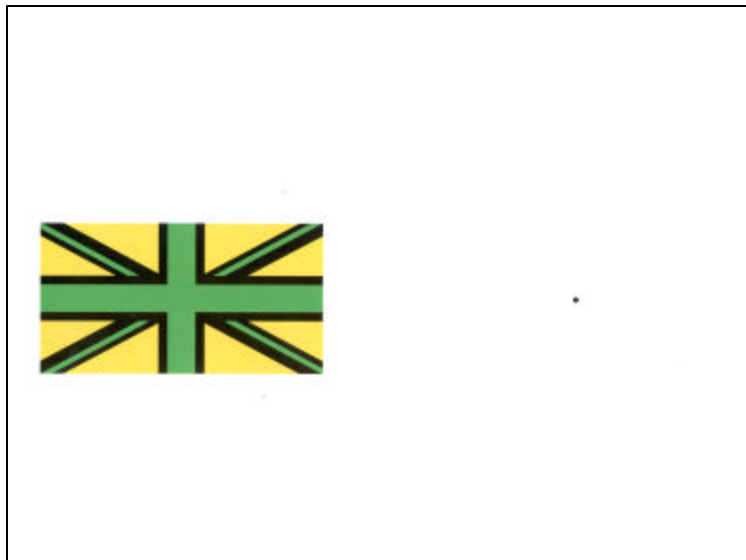
Lateral Inhibition + Opponent-Process =

- **Simultaneous *color* contrast** (similar to simultaneous lightness contrast, but with color, not lightness, affected by neighboring patch)

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- **Habituation:** Neurons that fire a lot over a long period of time "get tired."
- **Opponent-process circuitry plus habituation:**
 - Show stimulus A for a long time → A cells habituate
 - Show neutral stimulus → A cells habituated (below resting rate), B cells fresh (at resting rate)
 - $B > A$, so perceive neutral stimulus as B
- **Explains:**
 - Color aftereffects
 - Motion aftereffects



LOOK AT THIS IMAGE FOR 1 MINUTE . . .

WEEKLY WORLD NEWS

& a MIRACLE will happen!

Stare at the center of the image for one minute — and Jesus will appear before you HERE!

Directions for the faithful: Focus on the center of the miracle image for one full minute — then look at the white square on this page. A wonderful thing will happen!